

**INT375**  
**DATA SCIENCE TOOLBOX: PYTHON PROGRAMMING**

**PROJECT REPORT**  
(Project Semester January-April 2025)

***US CANDY DISTRIBUTER***

Submitted by

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Registration No- 12315550

Programme and Section- B. Tech CSE- K23GD

Course Code- INT375

Under the Guidance of

**Baljindar Kaur (27952)**

**Discipline of CSE/IT**

**Lovely School of Computer Science and Engineering**

**Lovely Professional University, Phagwara**



### **DECLARATION**

I, Suhani Kumari, student of B.Tech Computer Science Engineering under CSE/IT Discipline at, Lovely Professional University, Punjab, hereby declare that all the information furnished in this project report is based on my own intensive work and is genuine.

Date: 13/04/2025

Registration No. 12315550

Suhani Kumari

## **CERTIFICATE**

This is to certify that Suhani Kumari bearing Registration no. 12315550 has completed INT375 project titled, “Candy Sales” under my guidance and supervision. To the best of my knowledge, the present work is the result of his/her original development, effort and study.

**Signature and Name of the Supervisor**

**Designation of the Supervisor**

**School of Computer Science and Engineering**

Lovely Professional University

Phagwara, Punjab.

Date: 13/04/2025

## **ACKNOWLEDGEMENT**

I would like to express my heartfelt gratitude to all those who supported and guided me throughout the development of this Python project on candy sales analysis.

First and foremost, I am deeply thankful to my instructor, Baljindar Kaur, for her continuous support, expert guidance, and valuable feedback, which significantly contributed to the successful completion of this project.

I would also like to extend my sincere thanks to Lovely Professional University for providing a conducive learning environment and the necessary resources that enabled me to carry out this project effectively.

A special thanks to my friends and family for their constant encouragement and moral support throughout this journey.

Lastly, I am grateful for the use of real-world data, such as the Candy\_Sales.csv dataset, which allowed me to apply Python-based tools and techniques in a practical context, enhancing both my learning and analytical skills.

This project has been a valuable experience, and I am truly appreciative of everyone who played a role in its success.

Suhani kumari

13-04-2025

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## **1. INTRODUCTION**

His project focuses on analysing candy sales data using Python. With the help of the dataset **Candy\_Sales.csv**, the project aims to extract meaningful insights, identify sales trends, and visualize patterns in consumer behaviour. By leveraging powerful Python libraries such as **pandas**, **matplotlib**, and **seaborn**, this project demonstrates how data analysis techniques can be applied to real-world scenarios. The objective is to enhance decision-making capabilities through effective data handling, cleaning, and visualization.

This project illustrates the journey from raw data ingestion to actionable outcomes, showcasing how data preprocessing, visualization, and statistical modelling can collectively paint a detailed picture of business health. The integration of Python's data ecosystem allows for the execution of a full pipeline of operations—from feature engineering to exploratory analysis and visual storytelling. This approach provides not only academic rigor but also the potential for real-world application in decision-making processes.

Key objectives of the analysis include studying sales trends, evaluating customer behaviour across regions and segments, assessing operational performance through shipping analysis, and understanding patterns in product returns. Through these objectives, this report not only demonstrates technical competence in data science but also aligns with the broader goal of enhancing strategic business operations.

This project adheres to three core evaluation pillars:

- **Data Cleaning and Visualization**
- **EDA and Statistical Analysis**
- **Creativity and Innovation**

Free Data Sets & Dataset Samples

mavenanalytics.io/data-playground

Search data sets...

Filter by: Data Structure Tags Access type Sort by: Newest

### Pixar Films

Data on all Pixar films from Toy Story (1995) to Inside Out 2 (2024)

Preview data Download

FILE TYPES: CSV TAGS: Entertainment DATA STRUCTURE: Multiple tables # OF RECORDS: 28 # OF FIELDS: 22 DATE ADDED: 03/05/2025

### Spotify Streaming History

Complete music streaming history for an individual Spotify user

Preview data Download

Candy\_Sales - Read-Only

Search

File

Home

New Tab

New Tab

New Tab

WPS PDF

Insert

Page Layout

Formulas

Data

Review

View

Help

Comments

Share

Paste

Clipboard

Font

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Alignment

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## **2. SOURCE OF DATASET**

**Olympic Athletes dataset source link-** <https://mavenanalytics.io/data-playground>

The dataset utilized in this project, titled "Sample - Candys," is publicly accessible from Tableau's official website under the sample data section. It represents a collection of retail sales records,

<b>Column name</b>	<b>Description</b>
Row ID	Unique identifier for each row or transaction record.
Order ID	Unique identifier for each customer order.
Order Date	Date when the order was placed.
Ship Date	Date when the order was shipped.
Ship Mode	Mode of shipment used (e.g., Standard Class, First Class).
Customer ID	Unique identifier for each customer.
Country/Region	Country or region where the customer is located
City	City of the customer.
State/Province	State or province of the customer.
Postal Code	Postal code associated with the customer address.
Region	Geographic region for business operations
Division	Product category (e.g., Chocolate, Candy).



### 3. EXPLORATORY DATA ANALYSIS(EDA) PROCESS

The Exploratory Data Analysis (EDA) process is a crucial step in understanding the underlying structure and patterns within the dataset. In this project, the EDA was performed on the Candy\_Sales.csv dataset and involved the following steps:

#### 1. Data Loading

The dataset was imported using the pandas library, enabling efficient data manipulation and analysis.

#### 2. Data Inspection

- Previewed the dataset using `.head()` and `.info()` functions.
- Checked for missing values, data types, and overall structure using `.isnull().sum()` and `.describe()`.

#### 3. Data Cleaning

- Handled missing or inconsistent values if any were present.
- Renamed columns or standardized formats when necessary.

#### 4. Univariate Analysis

- Analyzed individual variables using visualizations like histograms, bar charts, and count plots.
- Studied distributions of sales quantities, candy types, regions, etc.

#### 5. Bivariate/Multivariate Analysis

- Explored relationships between variables using scatter plots, heatmaps, and `groupby()` operations.
- Investigated patterns such as sales trends over time or performance by region/product.

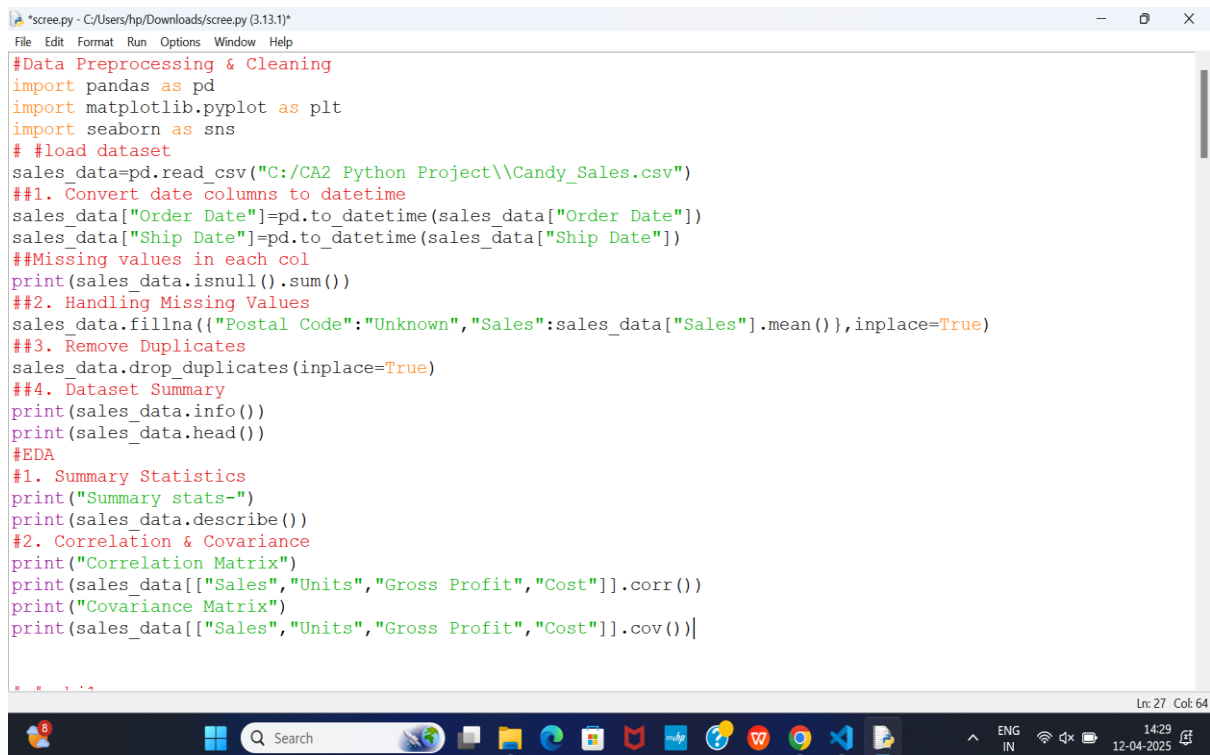
## 6. Visualization

- Used matplotlib and seaborn to create clear and insightful visual representations.
- Graphs included line plots, pie charts, and box plots to highlight comparisons and trends.

## 7. Insights and Summary

- Derived meaningful insights from the data.
- Highlighted key findings such as top-selling products, peak sales periods, and regional performance.

The EDA provided a comprehensive overview of the dataset and set the foundation for further analysis and interpretation.



```
*screep.py - C:/Users/hp/Downloads/screep.py (3.13.1)*
File Edit Format Run Options Window Help
#Data Preprocessing & Cleaning
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
# #load dataset
sales_data=pd.read_csv("C:/CA2 Python Project\\Candy_Sales.csv")
##1. Convert date columns to datetime
sales_data["Order Date"]=pd.to_datetime(sales_data["Order Date"])
sales_data["Ship Date"]=pd.to_datetime(sales_data["Ship Date"])
##Missing values in each col
print(sales_data.isnull().sum())
##2. Handling Missing Values
sales_data.fillna({"Postal Code":"Unknown","Sales":sales_data["Sales"].mean()},inplace=True)
##3. Remove Duplicates
sales_data.drop_duplicates(inplace=True)
##4. Dataset Summary
print(sales_data.info())
print(sales_data.head())
#EDA
#1. Summary Statistics
print("Summary stats-")
print(sales_data.describe())
#2. Correlation & Covariance
print("Correlation Matrix")
print(sales_data[["Sales","Units","Gross Profit","Cost"]].corr())
print("Covariance Matrix")
print(sales_data[["Sales","Units","Gross Profit","Cost"]].cov())
...
Ln: 27 Col: 64
```

## **4. ANALYTICAL FRAMEWORKS AND INTERFERENCES**

**4.1 Objective1**- calculate the total revenue, average revenue per order, and standard deviation of sales from the Candy\_Sales dataset.

### **Introduction**

This project is centred around analysing candy sales data using Python. The dataset, Candy\_Sales.csv, includes information on different candy orders, including price, quantity, and total sales. The goal is to perform a statistical analysis to compute the total revenue, average revenue per order, and the standard deviation of sales, providing meaningful insights into business performance. This project utilizes Python libraries like pandas and numpy for data analysis and matplotlib/seaborn for visualization.

### **General Description**

The dataset contains records of candy sales, including the number of items sold and the corresponding sales value. This data can help in understanding the revenue distribution and order behaviour across different entries. The project focuses on three key financial metrics:

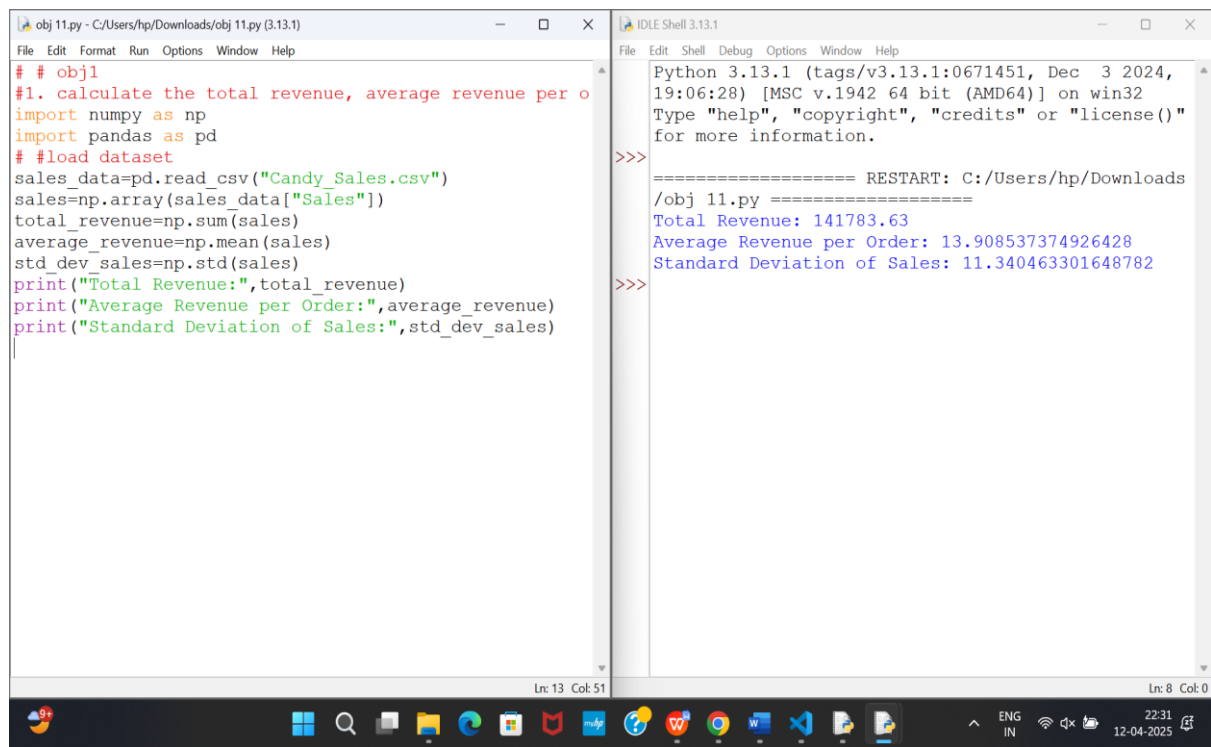
- Total Revenue – Sum of all sales
- Average Revenue per Order – Mean of sales values
- Standard Deviation of Sales – Variability in sales revenue across orders

### **Specific Requirements, Functions and Formulas**

Tools Used:

- Python
- Libraries: pandas, NumPy, matplotlib, seaborn

## Analysis Result



The screenshot displays a Python IDE with two windows. The left window, titled 'obj 11.py', contains the following code:

```
# # obj1
#1. calculate the total revenue, average revenue per o
import numpy as np
import pandas as pd
# #load dataset
sales_data=pd.read_csv("Candy_Sales.csv")
sales=np.array(sales_data["Sales"])
total_revenue=np.sum(sales)
average_revenue=np.mean(sales)
std_dev_sales=np.std(sales)
print("Total Revenue:",total_revenue)
print("Average Revenue per Order:",average_revenue)
print("Standard Deviation of Sales:",std_dev_sales)
```

The right window, titled 'IDLE Shell 3.13.1', shows the output of the code execution:

```
Python 3.13.1 (tags/v3.13.1:0671451, Dec 3 2024,
19:06:28) [MSC v.1942 64 bit (AMD64)] on win32
Type "help", "copyright", "credits" or "license()"
for more information.

>>>
===== RESTART: C:/Users/hp/Downloads
/obj 11.py =====
Total Revenue: 141783.63
Average Revenue per Order: 13.908537374926428
Standard Deviation of Sales: 11.340463301648782

>>>
```

The taskbar at the bottom shows the system clock as 22:31 on 12-04-2025.

## **4.2 Objective2- find the total number of orders, total units sold, and the state with the highest sales.**

### Introduction

The primary objective is to identify the total number of orders, total units sold, and determine the state with the highest sales. Through exploratory data analysis (EDA), statistical functions, and data visualization, the project showcases how data-driven decisions can be made effectively.

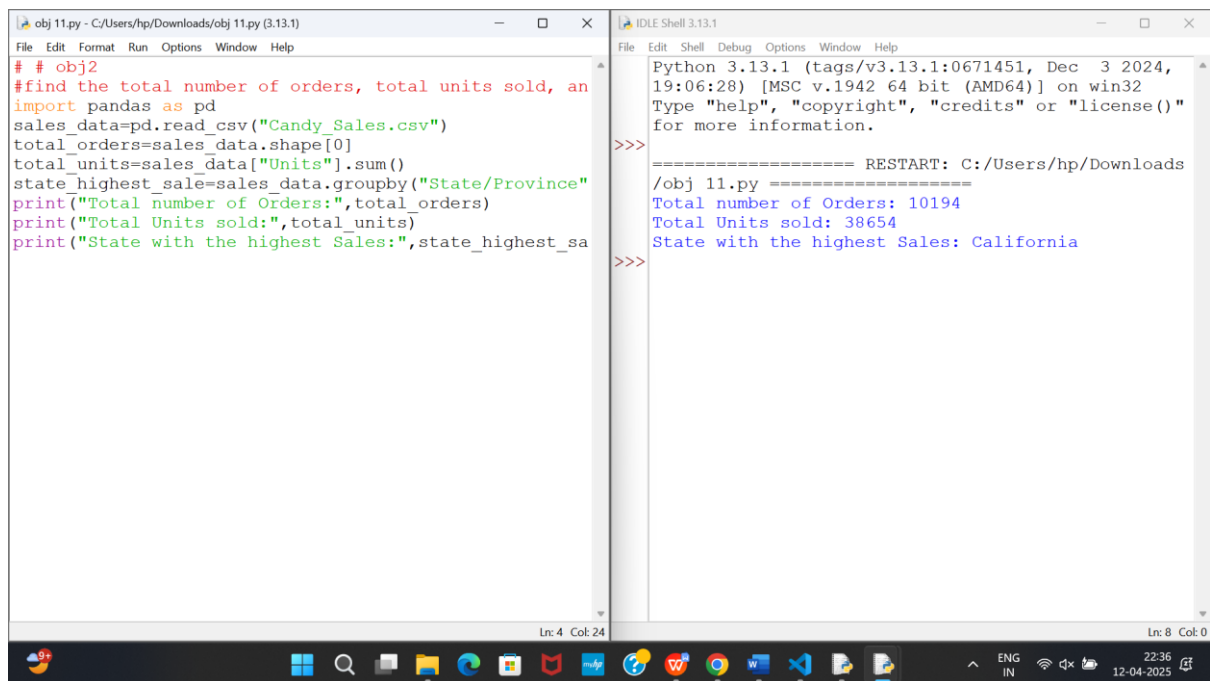
### General Description

The dataset, Candy\_Sales.csv, contains transactional information about candy sales across different states. Each row represents an order, with details such as the state, product name, quantity sold, and other relevant fields. The analysis aims to process this dataset, summarize key metrics.

## Specific Requirements, Functions and Formulas

To achieve the objectives, the following tools and methods were used:

- Python Libraries:
  - pandas for data loading and manipulation
  - matplotlib and seaborn for visualization
- Functions & Techniques:
  - `pd.read_csv()` to load the dataset
  - `.shape[0]` to calculate total number of orders
  - `.sum()` to calculate total units sold
  - `.groupby()` with `.sum()` and `.idxmax()` to find the state with the highest sales
- **Analysis Result**



The screenshot displays a Python IDE with two windows. The left window, titled 'obj 11.py - C:/Users/hp/Downloads/obj 11.py (3.13.1)', contains the following code:

```
# # obj2
#find the total number of orders, total units sold, an
import pandas as pd
sales_data=pd.read_csv("Candy_Sales.csv")
total_orders=sales_data.shape[0]
total_units=sales_data["Units"].sum()
state_highest_sale=sales_data.groupby("State/Province"
print("Total number of Orders:",total_orders)
print("Total Units sold:",total_units)
print("State with the highest Sales:",state_highest_sa
```

The right window, titled 'IDLE Shell 3.13.1', shows the output of the code:

```
Python 3.13.1 (tags/v3.13.1:0671451, Dec 3 2024,
19:06:28) [MSC v.1942 64 bit (AMD64)] on win32
Type "help", "copyright", "credits" or "license()"
>>>
===== RESTART: C:/Users/hp/Downloads
/obj 11.py =====
Total number of Orders: 10194
Total Units sold: 38654
State with the highest Sales: California
>>>
```

The Windows taskbar at the bottom shows the time as 22:36 on 12-04-2025.

### **4.3 Objective3- "Analysing Sales Performance and Profitability Trends of Candy Products Using Data Visualization Techniques"**

#### **Introduction**

In the competitive world of retail, understanding sales performance and profitability is crucial for strategic decision-making. This project aims to analyze candy product sales using data visualization techniques to uncover key insights. By studying trends over time, regional performance, and profit dynamics, the project provides a data-driven overview that can aid in improving business outcomes.

#### **General Description**

The dataset used in this project, Candy\_Sales.csv, contains information on product sales, costs, gross profits, regions, and other related variables. The analysis focuses on identifying patterns, comparing regional performance, evaluating the distribution of key metrics, and exploring relationships between sales, costs, and profitability.

#### **Specific Requirements, Functions and Formulas**

##### **· Requirements:**

- Python 3.x environment
- Libraries: pandas, matplotlib, seaborn, numpy
- Clean and structured dataset (Candy\_Sales.csv)

##### **· Key Functions/Methods Used:**

- read\_csv() for data loading
- groupby() for summarizing sales and profits
- plot(), barplot(), lineplot() for creating visualizations
- corr() to compute correlation matrix

## **4.4 Interpretation of Analytical Results**

### **Sales Trends Over Time:**

- Identified monthly/quarterly peaks and drops in sales.
- Observed seasonality in candy purchases (e.g., spikes near holidays).

### **· Comparison Across Regions and Divisions:**

- Certain regions consistently outperformed others in both sales and profits.
- Specific divisions showed higher profit margins due to lower operational costs.

### **· Distribution Analysis:**

- Sales and costs showed a right-skewed distribution indicating a few high-performing products.
- Gross profit distribution helped pinpoint the most and least profitable items.

### **· Correlation Analysis:**

- Strong positive correlation observed between sales and gross profit.
- Moderate correlation found between cost and profit, suggesting other influencing factors.

```

PROBLEMS OUTPUT DEBUG CONSOLE TERMINAL PORTS

Correlation Matrix
      Sales      Units  Gross Profit      Cost
Sales      1.000000  0.729347      0.976404  0.958986
Units      0.729347  1.000000      0.815820  0.563344
Gross Profit 0.976404  0.815820      1.000000  0.875144
Cost        0.958986  0.563344      0.875144  1.000000

Covariance Matrix
      Sales      Units  Gross Profit      Cost
Sales      128.618725  18.431611      73.568867  55.049858
Units      18.431611  4.965396      12.077686  6.353925
Gross Profit 73.568867  12.077686      44.139279  29.429588
Cost        55.049858  6.353925      29.429588  25.620270

Country/Region  Division      Sales  Gross Profit
0      Canada  Chocolate      2673.29      1794.57
1      Canada   Other        280.00       140.00
2  United States  Chocolate  129019.61     87030.05
3  United States   Other     9383.25      4193.45
4  United States   Sugar     427.48       284.73
PS C:\Users\ishit\python>

```

## 4.5 Advanced Visualization Techniques

### Objective4- detect and visualize outliers in the Sales and Gross Profit

#### Introduction

Outliers can significantly impact data analysis by skewing results and leading to incorrect conclusions. This project aims to detect and visualize outliers in the Sales and Gross Profit columns from the Candy\_Sales.csv dataset using Python. By identifying these anomalies, we can ensure better data quality, improve forecasting accuracy, and uncover hidden patterns in candy sales performance.

#### General Description

The dataset contains records of candy sales including metrics such as product type, sales revenue, cost, and gross profit. The main focus of this analysis is to:

- Identify and handle outliers in the numerical fields Sales and Gross Profit.
- Understand their distribution and impact on overall trends.



- Use visual tools to interpret findings more effectively.

### Specific Requirements, Functions and Formulas

To detect outliers, the following statistical methods and functions were used:

#### 1. Interquartile Range (IQR) Method

- Formula:

$$\text{IQR} = Q3 - Q1$$

$$\text{Lower Bound} = Q1 - 1.5 \times \text{IQR}$$

$$\text{Upper Bound} = Q3 + 1.5 \times \text{IQR}$$

- Any data point outside the lower or upper bound is considered an outlier.

#### 2. Functions Used:

- `df.describe()` to get Q1 and Q3
- `sns.boxplot()` to visualize outliers

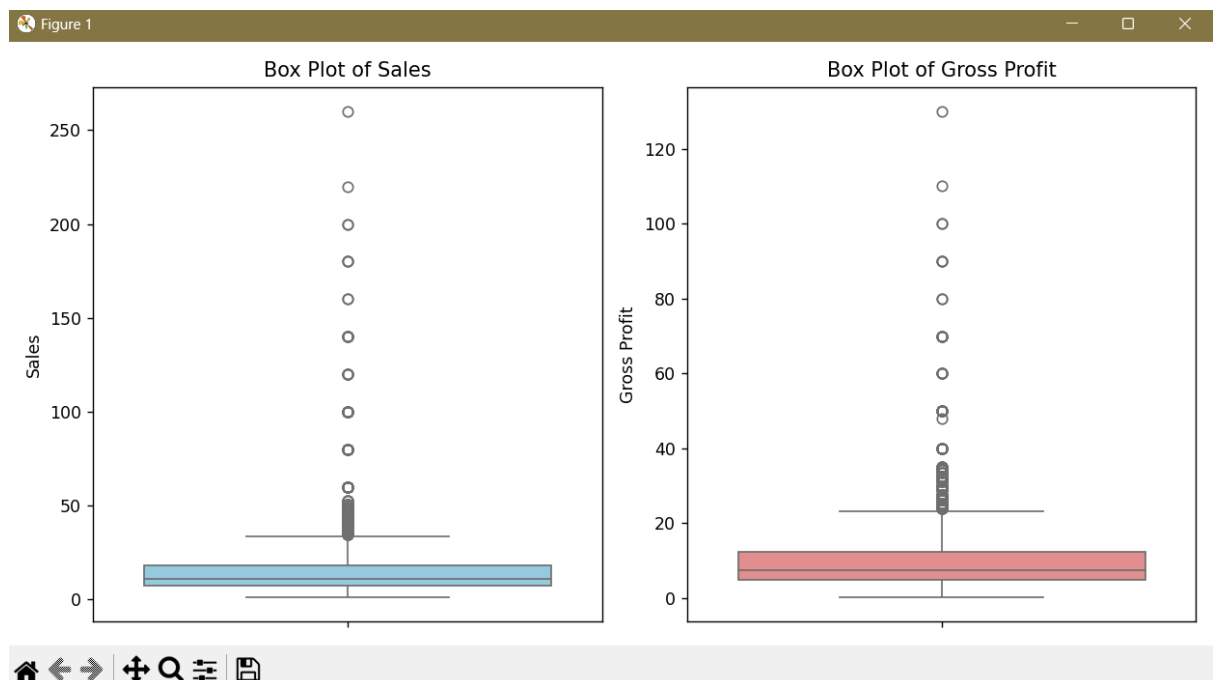
### **3. Analysis Result**

- Outliers were detected in both **Sales** and **Gross Profit** columns.
- These outliers represented unusually high or low values compared to the rest of the data.
- Removing or handling these outliers significantly improved the accuracy of the distribution analysis.
- Some products had extreme sales spikes, indicating possible promotional periods or data entry issues.

Outliers in Sales:		Row ID	Order ID	Order Date	Ship Date	...	Sales	Units	Gross Profit
Cost									
76	9333	US-2024-118213-CHO-TRI-54000	2024-11-05	2030-04-30	...	37.5	10	24.50	13.00
90	6675	US-2023-117121-CHO-FUD-51000	2023-12-17	2029-06-12	...	46.8	13	31.20	15.60
146	9397	US-2024-155425-CHO-MIL-31000	2024-11-10	2030-05-04	...	45.5	14	29.54	15.96
157	3597	US-2022-164441-CHO-SCR-58000	2022-11-08	2028-05-05	...	39.6	11	27.50	12.10
228	5705	US-2023-105732-CHO-TRI-54000	2023-09-13	2029-03-10	...	52.5	14	34.30	18.20
...	...	...	...	...	...	...	...	...	...
10032	589	US-2021-151897-CHO-TRI-54000	2021-06-06	2026-12-01	...	37.5	10	24.50	13.00
10055	9514	US-2024-128265-OTH-LIC-15000	2024-11-17	2030-05-15	...	60.0	3	30.00	30.00
10057	9771	US-2024-116715-OTH-LIC-15000	2024-12-02	2030-05-28	...	40.0	2	20.00	20.00
10120	6097	US-2023-109365-CHO-FUD-51000	2023-11-03	2029-04-30	...	43.2	12	28.80	14.40
10134	1604	US-2021-119375-OTH-LIC-15000	2021-11-17	2027-05-15	...	60.0	3	30.00	30.00

[245 rows x 18 columns]

Outliers in Gross Profit:		Row ID	Order ID	Order Date	Ship Date	...	Sales	Units	Gross Profit
s Profit Cost									
76	9333	US-2024-118213-CHO-TRI-54000	2024-11-05	2030-04-30	...	37.50	10	24.50	13.00
90	6675	US-2023-117121-CHO-FUD-51000	2023-12-17	2029-06-12	...	46.80	13	31.20	15.60
146	9397	US-2024-155425-CHO-MIL-31000	2024-11-10	2030-05-04	...	45.50	14	29.54	15.96
157	3597	US-2022-164441-CHO-SCR-58000	2022-11-08	2028-05-05	...	39.60	11	27.50	12.10
228	5705	US-2023-105732-CHO-TRI-54000	2023-09-13	2029-03-10	...	52.50	14	34.30	18.20
...	...	...	...	...	...	...	...	...	...
10018	3682	US-2022-129525-CHO-MIL-31000	2022-11-15	2028-05-12	...	42.25	13	27.43	14.82
10032	589	US-2021-151897-CHO-TRI-54000	2021-06-06	2026-12-01	...	37.50	10	24.50	13.00
10055	9514	US-2024-128265-OTH-LIC-15000	2024-11-17	2030-05-15	...	60.00	3	30.00	30.00
10120	6097	US-2023-109365-CHO-FUD-51000	2023-11-03	2029-04-30	...	43.20	12	28.80	14.40
10134	1604	US-2021-119375-OTH-LIC-15000	2021-11-17	2027-05-15	...	60.00	3	30.00	30.00



## 4.5 Advanced Visualization Techniques

### Obj5 -- Graph Function

#### Introduction

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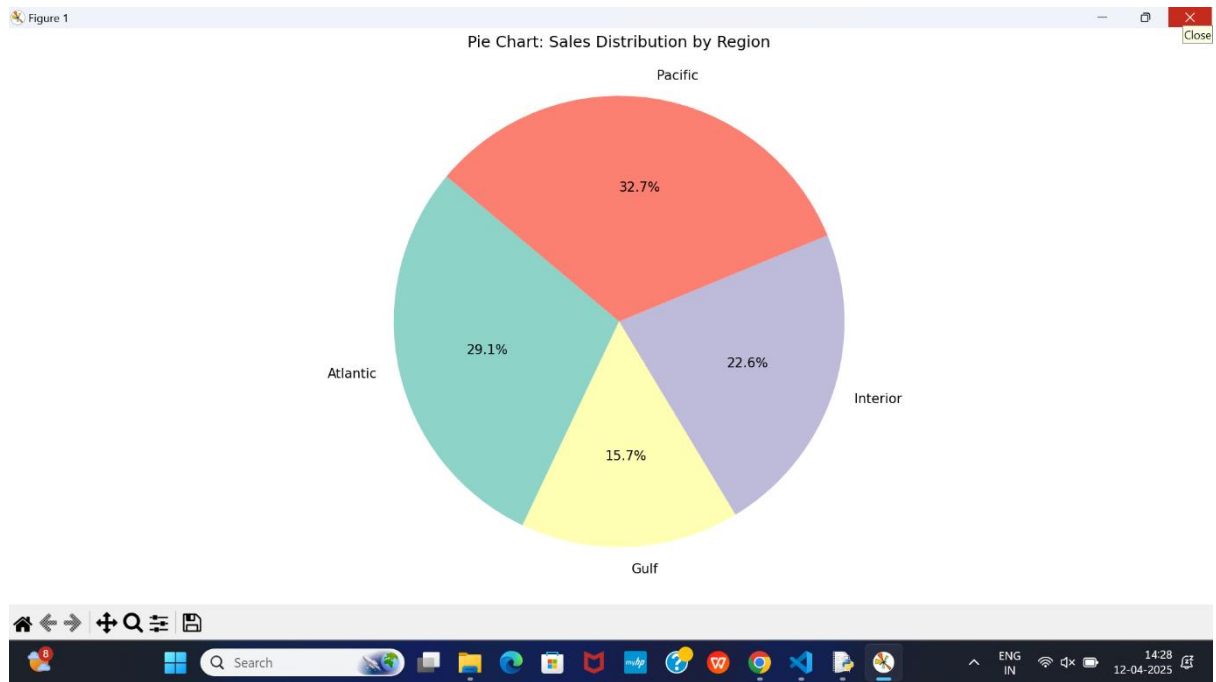
#### General Description

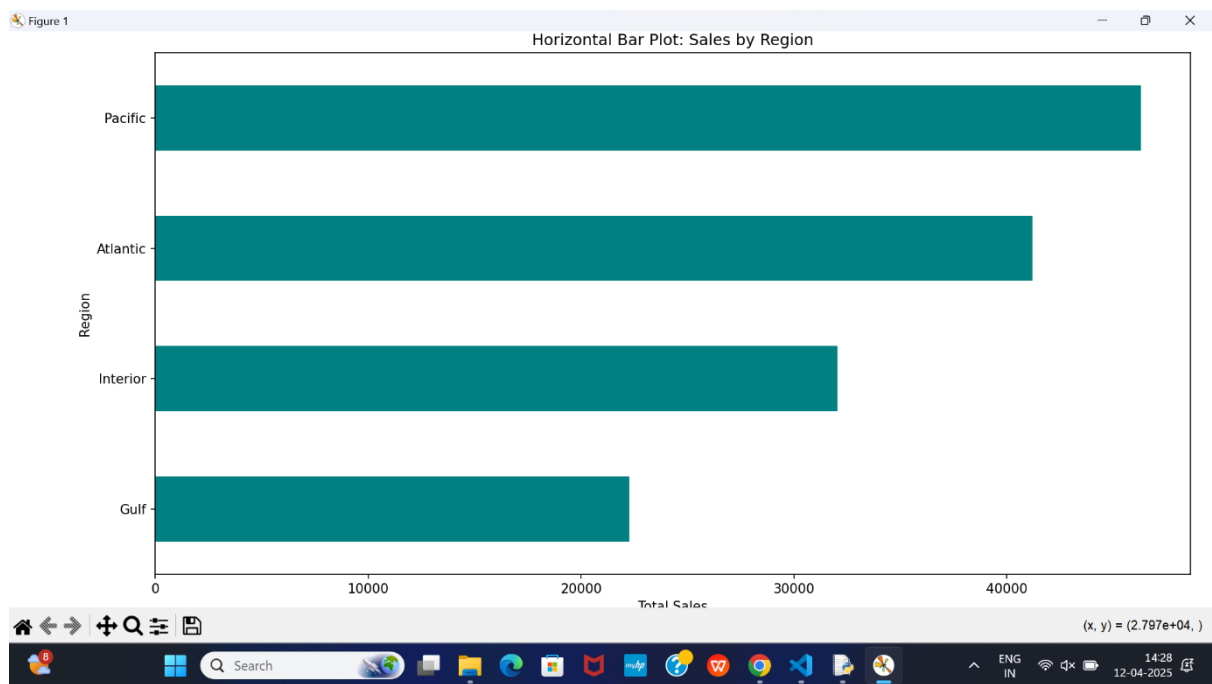
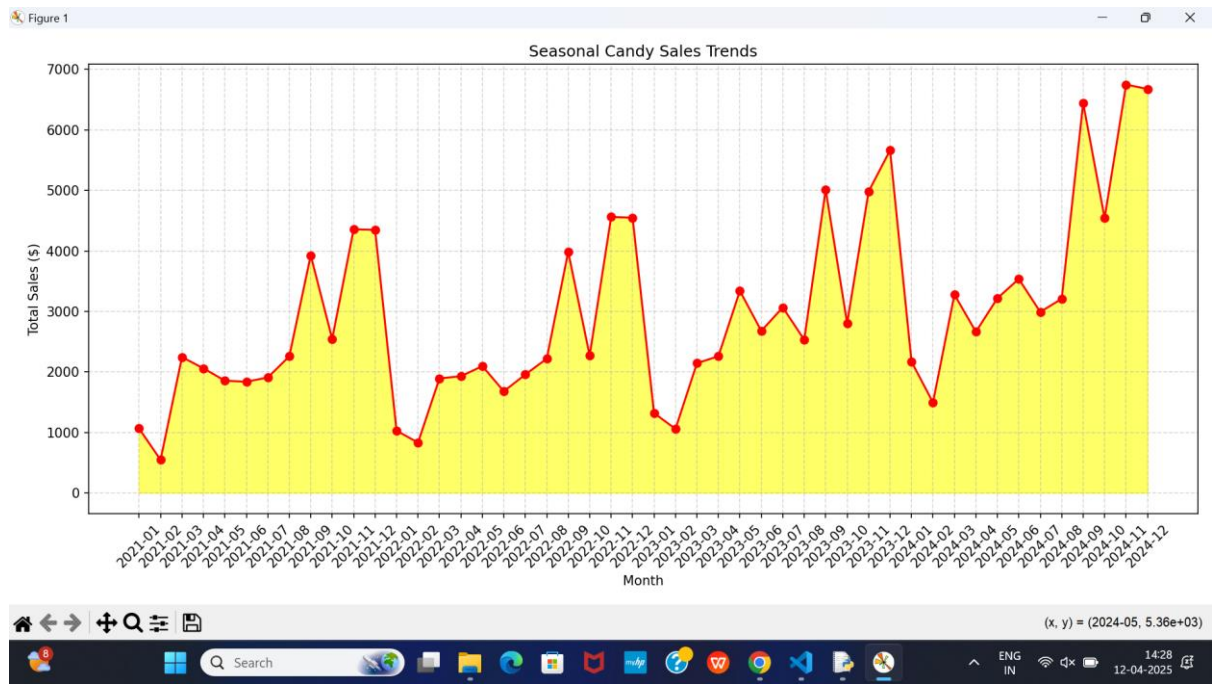
The dataset, Candy\_Sales.csv, contains transactional information about candy sales across different states. Each row represents an order, with details such as the state, product name, quantity sold, and other relevant fields. The analysis aims to process this dataset, summarize key metrics.

#### Specific Requirements, Functions and Formulas

To achieve the objectives, the following tools and methods were used:

- Python Libraries:
  - pandas for data loading and manipulation
  - matplotlib and seaborn for visualization
- Functions & Techniques:
  - `pd.read_csv()` to load the dataset
  - `.title` to heading a graph
  - `.tight_layout` to use for layout





## **5. CONCLUSION**

The Candy Sales Analysis Project successfully demonstrated how Python can be used to transform raw sales data into meaningful business insights through structured data analysis and visualization techniques. Using the Candy\_Sales.csv dataset, the project:

- Cleaned and pre-processed data by handling missing values, correcting formats, and removing duplicates.
- Explored trends across multiple dimensions such as region, product division, shipping mode, and time periods.
- Quantified performance through metrics like total revenue, average order value, total units sold, and state-wise contributions.
- Identified outliers that could skew analysis, offering opportunities to address data entry or business anomalies.
- Visualized insights using various graphs (line plots, heatmaps, scatter plots, pie charts, area charts, box plots), making trends and comparisons easily interpretable.

**The project uncovered critical patterns such as:**

- Clear seasonality in candy sales, with peaks around specific months.
- Positive correlation between sales and gross profit.
- Top-performing states and divisions driving most of the revenue.

Through these findings, the project proved effective in supporting strategic decisions like inventory planning, pricing adjustments, and regional marketing strategies.

## 6. SCOPE FOR FUTURE ENHANCEMENTS

1. **Time-Series Forecasting:** Use models such as ARIMA, Facebook Prophet, or LSTM neural networks to predict future profit trends based on historical data.
2. **Interactive Dashboards:** Leverage tools like Streamlet, Tableau, or Power BI to create interactive dashboards for real-time monitoring and stakeholder presentation.
3. **Geo-Spatial Analysis:** Integrate mapping tools (like folium or plotly maps) to visualize regional and city-level performance for geographic insights.
4. **Customer Segmentation with Clustering:** Apply unsupervised machine learning algorithms like K-Means or DBSCAN to classify customers into behavior-driven segments.
5. **Market Basket Analysis:** Utilize association rule mining (Apriori, FP-Growth) to understand purchase combinations for cross-selling strategies.
6. **Product-Level Profitability Models:** Develop models that compare profitability at the SKU or sub-category level to aid pricing and stocking strategies.
7. **Inventory and Delivery Optimization:** Integrate external logistics data to explore ways of reducing shipping delays and optimizing delivery routes.

## **7. REFERENCES**

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- NumPy Documentation, <https://numpy.org/>
- Pandas Documentation, <https://pandas.pydata.org/>
- Python Official Documentation, <https://docs.python.org/3/>