

# Sales Data Analysis Project

## 1. Project Overview and Objectives

### Project Overview

The Sales Data Analysis project is a Python-based data analytics application that processes a structured sales dataset to extract meaningful business insights. Using the pandas library, the project performs data cleaning, validation, analysis, and reporting to support data-driven decision-making.

### Objectives

- Load and preprocess real-world sales data
- Handle missing values and duplicate records
- Calculate key business metrics such as revenue and sales volume
- Identify best-selling products and top-performing regions
- Generate a clean, readable sales analysis report

## 2. Setup and Installation Instructions

### Prerequisites

- Python 3.8 or higher installed on the system
- Pip (Python package manager)

### Installation Steps

1. Download or clone the project repository from GitHub
2. Install required dependencies
3. Navigate to the project directory
4. Run the application using:

```
pip install -r requirements.txt
```

```
python sales_analysis.py
```

### 3. Code Structure Explanation

```
sales-data-analysis/  
|  
├── sales_analysis.py    # Main analysis script  
├── sales_data.csv       # Input dataset  
├── analysis_report.md   # Analysis results and insights  
├── requirements.txt     # Python dependencies  
└── screenshots  
    └── output.png      #Result
```

#### File Descriptions

- sales\_analysis.py  
Contains all logic for loading, cleaning, analyzing, and reporting data.  
Uses modular functions for readability and maintainability.
- sales\_data.csv  
Source dataset containing sales transactions.
- analysis\_report.md  
Summarized findings and business insights.
- requirements.txt  
Lists external libraries required to run the project.
- screenshot.png  
A screenshot demonstrating the successful execution of the application and sample output.

### 4. Screenshots of Working Application

The screenshot.png file included in the project directory shows:

- The dataset was loaded and explored correctly
- Data types were identified properly
- Sales metrics were computed successfully

- The final report was formatted and printed as expected

This visual proof confirms that the application runs correctly and meets functional requirements.

```
Data Shape: (100, 7)

Columns: ['Date', 'Product', 'Quantity', 'Price', 'Customer_ID', 'Region', 'Total_Sales']

Data Types: Date          object
Product       object
Quantity      int64
Price         int64
Customer_ID   object
Region        object
Total_Sales   int64
dtype: object

-----Sales Report-----

Total Revenue: ₹12,365,048.00

Best Selling Product: Laptop

Total Quantity Sold: 478

Average Order Value: ₹123,650.48

Top-performing Region: North

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```

5. Technical Requirements Fulfillment Explanation

Category	Technical Requirement	How It Was Implemented
Algorithm	Data loading and preprocessing flow	A sequential input–process–output workflow was followed to load the CSV file, validate required columns, clean the dataset, and prepare it for analysis.
Algorithm	Data cleaning logic	Missing numeric values were handled using fillna(), duplicate rows were removed using drop_duplicates(), and invalid dates were safely handled using pd.to_datetime() with error coercion.

Category	Technical Requirement	How It Was Implemented
Algorithm	Sales metric calculations	Aggregation functions such as <code>sum()</code> , <code>mean()</code> , and <code>groupby()</code> were used to compute total revenue, average order value, and product- and region-wise performance.
Algorithm	Best performer identification	Grouped aggregation with <code>idxmax()</code> was applied to identify the best-selling product and top-performing region based on total sales.
Data Structure	Tabular data handling	A pandas DataFrame was used to store and manipulate structured sales data efficiently.
Data Structure	Metric storage	Calculated metrics were stored in variables and returned as a tuple for organized access and reporting.
Architecture	Modular program design	Separate functions were created for data loading, validation, cleaning, analysis, and report generation to improve readability and maintainability.
Architecture	Configuration management	Constants such as file paths and date formats were defined at the top of the program to centralize configuration.
Architecture	Single-file execution	All program logic was contained within <code>sales_analysis.py</code> , with a <code>main()</code> function controlling execution flow.
Output	Formatted reporting	A dedicated report function was implemented to display results in a clean, user-friendly, and well-formatted console output.

## 6. Learning Outcomes

Through this project, the following learning outcomes were achieved:

- Gained hands-on experience in loading, cleaning, and analyzing structured data using pandas.
- Learned how to handle missing values, duplicates, and data type conversions effectively.

- Developed the ability to compute key business metrics such as revenue, averages, and best-performing categories.
- Improved understanding of modular program design and clean code structure in Python.

## **7. Conclusion**

This project analyzes sales data using Python and pandas to extract key business insights such as revenue, best-selling products, and regional performance, emphasizing the importance of data cleaning and structured analysis.