



AGRICULTURE PRODUCTIVITY AND WATER FOOTPRINT ANALYSIS



SUMMER INTERNSHIP REPORT

Submitted by

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613022205044

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of

BACHELOR OF TECHNOLOGY

in

INFORMATION TECHNOLOGY

**VIVEKANANDHA COLLEGE OF TECHNOLOGY FOR WOMEN
ELAYAMPALAYAM, TIRUCHENGODE-637205**

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**VIVEKANANDHA
COLLEGE OF TECHNOLOGY FOR WOMEN**

(Approved by AICTE, New Delhi, Affiliated to Anna University, Chennai.)

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BONAFIDE CERTIFICATE

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Summer Internship Viva-Voce Examination held on _____

INTERNAL EXAMINER

EXTERNAL EXAMINER

COMPANY PROFILE



1.COMPANY OVERVIEW:

Name of the Company: Profenaa Technologies & Automation

Industry : Engineering Services, Industrial Automation, and Technical Consultancy

Location: Ramanathapuram Branch, Coimbatore, Tamil Nadu

Mission Statement: Empowering growth through innovative engineering solutions and expert software training, with real-time project experience and client collaboration. Profenaa bridges the gap between education and industry by equipping students and professionals with the technical skills and on-job exposure required to excel in the modern engineering world.

Core Products/Services:

- Mechanical, Civil, Electrical, and Electronics Engineering Projects
- Industrial Automation – PLC, SCADA, and Robotics Solutions
- CAD/CAM/CAE Design and Simulation Services
- Technical Training and Skill Development
- Job Placement and Industry Collaboration Consultancy

COMPANY SIZE AND STRUCTURE

Employee Count:

Profenaa technologies & Automation operates as a mid-sized organization, typically comprising **50–100 employees** across its engineering, training, and consulting divisions.

Departments and Teams:

1. **Engineering Services Division:** Focuses on delivering customized industrial projects in Mechanical, Civil, and Electrical domains.
2. **Automation & Robotics Division:** Handles industrial automation, PLC, SCADA, IoT, and smart manufacturing solutions.
3. **Design and Product Development Team:** Specializes in CAD/CAM/CAE, 3D printing, and simulation-based design.
4. **Training & Placement Division:** Provides real-time industrial training and connects students with industry partners.
5. **Business Development & Client Relations:** Works with partner companies to provide engineering solutions and recruitment support.

MARKET POSITION AND COMPETITORS

Market Position:

- Profenaa Technologies & Automation is a reputed **engineering and industrial automation firm** based in Coimbatore, offering a one-stop platform for project design, product development, and industry training. The company collaborates with major industries such as **TAFE, General Motors, and Flowserve Corporation** and is recognized for combining **practical engineering expertise** with **career development opportunities** for students.

Key Competitors:

1. Caresoft Global
2. Axiscades
3. Skyappz Software India Pvt. Ltd.
4. Altran Technologies

FUTURE PROSPECTS AND GROWTH OPPORTUNITIES

Profenaa Technologies aims to expand its reach and technological capabilities in the coming years by:

- **Integrating Industry 4.0 and Smart Manufacturing:** Implementing IoT, robotics, and predictive maintenance systems in client projects.
- **Expanding Automation and Robotics Training:** Introducing hands-on PLC/SCADA and Industrial Robotics courses for students.
- **Enhancing Academic Collaborations:** Partnering with more colleges and universities to provide real-time industrial exposure.
- **Advancing Research in 3D Printing and Simulation:** Focusing on additive manufacturing and FEA/CFD analysis to support product innovation.

RECENT NEWS AND DEVELOPMENTS

1. Industry-Academia Collaboration:

Profenaa conducted multiple career guidance sessions **and** industrial training programs **for engineering students, helping them understand current industry trends and technologies like robotics, additive manufacturing, and automation.**

2. Successful Client Projects:

Completed projects for companies such as Fabtech Industries, Ashjoe Cold Chain, **and** Arrow Tech Private Limited (Singapore) **in fields like** machine design, digital marketing, and web development **using modern CAD and full-stack tools.**

CORPORATE SOCIAL RESPONSIBILITY (CSR)

- **Educational Responsibility:** Provides affordable and practical engineering training for students from various disciplines.
- **Community Development:** Supports employability initiatives by connecting graduates with industrial placements.
- **Environmental Awareness:** Encourages the adoption of sustainable engineering practices and digital workflows to reduce material waste.

PROFESSIONAL DEVELOPMENT OPPORTUNITIES

- **Training and Mentorship:** Offers structured, hands-on training using industry-standard tools such as **AutoCAD, SolidWorks, CATIA, ANSYS, and MATLAB.**
- **Real-Time Project Exposure:** Interns work directly on live industrial projects under expert supervision.
- **Career Growth Support:** Provides job placement assistance with core companies in **mechanical, civil, ECE, and IT sectors.**

CHALLENGES FACED BY PROFENAA TECHNOLOGIES IN COIMBATORE

1. **Talent Retention:** Skilled graduates often migrate to larger metros, creating a challenge for retaining local talent in Coimbatore.
2. **Keeping Pace with Technological Changes:** Rapid developments in automation and Industry 4.0 demand continuous upskilling of trainers and engineers.
3. **Awareness Among Students:** Many engineering students lack practical exposure; Profenaa is addressing this through job-integrated learning programs.

CONTACT INFORMATION

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[illegible]

| NAME | ATTENDANCE | ATTENDANCE | ATTENDANCE | ATTENDANCE |
|---------------|------------|------------|------------|--------------------|
| | WEEK 1 | WEEK 2 | WEEK 3 | WEEK 4 |
| | PYTHON | SQL | EXCEL | POWER BI & TABLEAU |
| ABINIYA A | PRESENT | PRESENT | PRESENT | PRESENT |
| CHARU S | PRESENT | PRESENT | PRESENT | PRESENT |
| DEEPIKA G | PRESENT | PRESENT | PRESENT | PRESENT |
| DHARSHANA V | PRESENT | PRESENT | PRESENT | PRESENT |
| HARINI K | PRESENT | PRESENT | PRESENT | PRESENT |
| MANISHA I | PRESENT | PRESENT | PRESENT | PRESENT |
| NAVITH A | PRESENT | PRESENT | PRESENT | PRESENT |
| PRIYA B | PRESENT | PRESENT | PRESENT | PRESENT |
| ROOHI F | PRESENT | PRESENT | PRESENT | PRESENT |
| SALMA V | PRESENT | PRESENT | PRESENT | PRESENT |
| SALVA M | PRESENT | PRESENT | PRESENT | PRESENT |
| SANGEETHA N | PRESENT | PRESENT | PRESENT | PRESENT |
| SANTHIYA S | PRESENT | PRESENT | PRESENT | PRESENT |
| SIVA M | PRESENT | PRESENT | PRESENT | PRESENT |
| TARIQ F | PRESENT | PRESENT | PRESENT | PRESENT |
| WILSON B | PRESENT | PRESENT | PRESENT | PRESENT |
| ZIYA FATIMA F | PRESENT | PRESENT | PRESENT | PRESENT |

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ACKNOWLEDGEMENT

First of all ,we thank the almighty for being with me all ways for giving me due courage, insight and wisdom to take up and complete this Internship study successfully in time.

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ABSTRACT

The **Data Analytics Management System (DAMS)** is an integrated web-based platform designed to streamline the process of managing, analyzing, and visualizing large datasets efficiently. Using a **full stack development and data-driven approach**, the system combines both front-end and back-end technologies to deliver an interactive and user-friendly solution for data analysts, business professionals, and organizations. The platform enables users to **upload, process, and visualize datasets**, while ensuring secure data storage and real-time analytical insights.

The **front-end** of the system is developed using modern web technologies such as **HTML, CSS, and JavaScript**, along with frameworks like **React.js or Angular**, to create an intuitive and responsive user interface for interactive data visualization. The **back-end** leverages technologies such as **Python (Flask/Django), Node.js, Express.js, and MongoDB/MySQL**, providing powerful data handling, secure authentication, and smooth integration between the front-end and database.

Additionally, the system incorporates advanced features such as **data cleaning, statistical analysis, predictive modeling, and interactive dashboards** using tools like **Pandas, NumPy, Matplotlib, Power BI, or Tableau**. These capabilities help users derive meaningful insights and make data-driven decisions effectively.

The Data Analytics Management System supports organizations in **optimizing decision-making processes**, reducing manual effort in data analysis, and improving business intelligence capabilities. It also minimizes paper-based reporting and promotes digital transformation by providing **automated report generation and real-time dashboards**.

Essentially, this software includes **three core modules**:

1. **Data Collection and Preprocessing Module** – for importing, cleaning, and preparing raw datasets.
2. **Data Analysis Module** – for performing statistical, descriptive, and predictive analysis.
3. **Data Visualization Module** – for creating dashboards, charts, and visual reports to communicate insights effectively.

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CHAPTER 1

INTRODUCTION

1.1 About Internship

The “**Data Analytics**” internship at **Profenna Technologies & Automation, Coimbatore**, is a comprehensive program aimed at providing a strong foundation in data analytics and visualization techniques. The internship focuses on developing analytical skills through tools such as **Python, SQL, Excel, Power BI, and Tableau**, helping interns transform raw data into meaningful insights. Participants n data analytics workflow — from **data collection, cleaning, and processing to visualization and reporting**.

- To gain hands-on experience with **data analytics tools** such as Python, SQL, Excel, Power BI, and Tableau.
- To evaluate the role of data analytics in improving **business decision-making and operational efficiency**.
- To identify and interpret **key performance indicators (KPIs)** through real-time dashboards.
- To enhance skills in **data visualization and storytelling**, making analytical results understandable for decision-makers.

1.3 Scope of the Study

The scope of this internship encompasses a variety of **data analytics techniques and tools** essential for handling and interpreting business data effectively:

- Focuses on **data preprocessing, analysis, and visualization** across multiple platforms such as Excel, SQL, Power BI, and Tableau.
- Includes the study of **data collection methods, database management, and query optimization** using SQL.

- Covers **data visualization and dashboard building** to present insights clearly and interactively.
- Involves **real-world business datasets**, enabling interns to apply analytics techniques to practical scenarios.
- Explores the use of analytics in **improving organizational efficiency** and supporting **data-driven decision-making**.

1.4 Limitations of the Study

While the internship provided valuable hands-on experience, certain limitations were encountered during the study:

- The analysis was conducted using limited datasets, which may not represent all real-world business cases.
- Time constraints restricted the depth of exploration in advanced topics like predictive analytics and machine learning.
- Some tools and datasets had restricted access, limiting exposure to enterprise-level data platforms.
- Dashboard results depended on the accuracy and completeness of the available data sources.

1.5 Duration of the Study

The internship spanned a period of **27th June 2025 to 27th July 2025**, providing one full month of structured training and practical project work. The program was divided into four weekly learning phases:

- **Week 1:** Python for Data Analytics
- **Week 2:** SQL and Excel for Data Handling and Reporting
- **Week 3:** Power BI for Business Intelligence and Dashboard Creation
- **Week 4:** Tableau for Advanced Visualization and Insights

CHAPTER 2

WORKFLOW AND METHODOLOGY

2.1 Background: Context and Motivation for the Project

In the digital era, data has become one of the most valuable assets for businesses and organizations. Every transaction, process, and activity generates data that, when properly analyzed, provides meaningful insights into performance, efficiency, and growth opportunities. The **Data Analytics Internship Program at Profenaa Technologies & Automation, Coimbatore**, was initiated to train aspiring data professionals in utilizing these datasets for actionable business intelligence.

The background of this project lies in the increasing demand for **data-literate professionals** who can extract, interpret, and visualize data using analytical tools. Through this internship, students are exposed to **practical business problems** that require data-based solutions, allowing them to apply analytical techniques to real-world scenarios.

The motivation behind undertaking this internship was to gain **hands-on experience in data analytics tools and techniques**, which are essential in modern organizations. By mastering tools such as **Python, SQL, Excel, Power BI, and Tableau**, the internship aims to equip participants with the ability to handle large datasets, generate dashboards, and present findings effectively.

Furthermore, the project emphasizes understanding the **entire data lifecycle** — from data collection and preprocessing to visualization and reporting. This exposure allows interns to comprehend how data analytics is used in decision-making, forecasting, and evaluating business outcomes. Ultimately, the internship motivates participants to adopt **data-driven thinking** and enhance their problem-solving capabilities.

2.2 Objective: Specific Goals and Objectives of the Project

The main objective of this internship is to apply **data analytics principles** to transform raw data into meaningful insights that can guide decision-making. The project focuses on gaining practical exposure to **data cleaning, analysis, visualization, and interpretation** using a combination of analytical tools and programming languages.

The specific objectives include:

- To develop a clear understanding of the **data analytics workflow**, including data gathering, preprocessing, and transformation.
- To build technical proficiency in **Python (Pandas, NumPy)** for data manipulation and analysis.
- To learn and implement **SQL queries** for database management, data extraction, and aggregation of large datasets.
- To analyze data trends and patterns using **Excel** functions, formulas, and pivot tables.
- To create **interactive dashboards** using **Power BI** and **Tableau**, representing insights in a visually engaging manner.
- To enhance analytical thinking by applying **real-time case studies** and datasets to simulate business scenarios.
- To evaluate data visualization techniques that improve communication of insights to non-technical stakeholders.
- To gain exposure to **data ethics and privacy practices**, ensuring that all analyses adhere to confidentiality and integrity standards.

This internship's ultimate goal is to develop both the **technical and analytical skillsets** necessary for a successful career in data analytics and business intelligence.

2.3 Scope: The Scope and Boundaries of the Project

The scope of this Data Analytics project encompasses all essential stages of the analytical process, focusing on the conversion of raw datasets into useful, business-oriented insights. The internship covers a diverse range of technologies and methodologies that are integral to a data analytics environment.

SCOPE OF STUDY

- To understand **data collection techniques**, including importing datasets from Excel, SQL databases, and APIs.
- To perform **data preprocessing** — cleaning, handling missing values, normalization, and transformation for consistency.
- To apply **statistical and exploratory analysis** using Python libraries like Pandas and NumPy.
- To utilize **SQL** for database querying, relational joins, aggregations, and analytical reporting.
- To use **Excel** for tabular analysis, chart creation, and pivot table summarization.
- To create **interactive dashboards** in Power BI and Tableau that visualize KPIs, performance metrics, and time-based trends.
- To interpret findings and draw **business insights** from dashboards, focusing on decision support and performance improvement.

Boundaries of the Study:

- The study primarily focuses on **descriptive and diagnostic analytics**, excluding advanced topics like machine learning or predictive analytics due to time constraints.
- The project uses **limited sample datasets** provided by the company or open data repositories, which may not represent the complexity of enterprise-level data.
- The analysis emphasizes **visual reporting and dashboard creation**, rather than algorithmic modeling.
- Resource and tool access is limited to open-source or licensed academic tools available during the internship duration.

By maintaining this scope, the internship ensures that participants develop strong analytical fundamentals while remaining focused on tools and techniques relevant to industry requirements.

2.4 Methodology: Overview of the Approach and Methodologies Used

The **Data Analytics Methodology** adopted in this project follows a systematic, step-by-step approach designed to simulate an end-to-end real-world analytics process. Each phase involves

specific activities that collectively contribute to the creation of reliable insights and visual dashboards.

Phase 1 – Data Collection

The first stage involves identifying relevant datasets and gathering them from appropriate sources. Data was collected from **Excel sheets, SQL databases, and public repositories** related to performance tracking, sales, and student analytics. The datasets were then imported into Python and Power BI for further processing.

- Tools used: Excel, SQL Server, Python (Pandas).
- Output: Raw dataset files in .csv and .xlsx formats.

Phase 2 – Data Cleaning and Preprocessing

Raw data often contains missing values, duplicates, and inconsistencies. This phase involved performing **data validation, transformation, and formatting** to ensure data quality.

- Handled null values, corrected data types, and standardized text fields.
- Used **Excel Power Query** and **Python Pandas** for cleaning operations.
- Ensured uniformity across columns, such as consistent date formats and naming conventions.

Phase 3 – Data Analysis

In this phase, analytical techniques were applied to identify patterns, correlations, and key insights from the cleaned data.

- Performed **Exploratory Data Analysis (EDA)** using Python, SQL, and Excel formulas.
- Generated summaries, pivot tables, and grouped aggregations to study data behavior.
- Applied statistical measures (mean, median, mode, variance) for descriptive analysis.

Phase 4 – Data Visualization

The cleaned and analyzed data was visualized using **Power BI** and **Tableau** to create meaningful dashboards.

- Developed charts such as bar graphs, pie charts, and line plots to represent key metrics.
- Designed **interactive dashboards** with slicers and filters for real-time analysis.
- Focused on visual storytelling — transforming numerical data into business-friendly insights.

Phase 5 – Reporting and Interpretation

The insights derived from visualizations were compiled into a comprehensive analytical report.

- Reports were created using Power BI, Tableau, and Excel.
- Key findings were summarized with visual aids for easier interpretation.
- Each visualization was linked to an actionable business insight (e.g., performance trends, attendance ratios, growth patterns).

Phase 6 – Evaluation and Presentation

The final step involved evaluating the effectiveness of the analytical process.

- Reviewed dashboards with mentors and peers for feedback.
- Assessed the clarity, interactivity, and interpretability of dashboards.
- Suggested improvements for scalability and integration with larger datasets in future work.

CHAPTER 3

SYSTEM STUDY

3.1 Existing Solutions

Before implementing the Data Analytics Management System (DAMS), most organizations and institutions relied on **traditional, manual, or semi-automated methods** to manage and interpret their data. These methods were often based on spreadsheets, static reports, and disconnected tools, making the analytical process time-consuming and prone to human error.

In the existing system, data was usually collected and maintained through **Excel sheets or basic databases**, with limited analytical and visualization capabilities. Each department or team often worked in isolation, leading to **data silos** and inconsistencies in reporting. The lack of integrated dashboards made it difficult to gain real-time insights or identify performance trends across different dimensions.

Some of the major challenges identified in the existing systems include:

1. **Manual Data Handling:**

Data entry, cleaning, and reporting were primarily manual, which increased the risk of inaccuracies and duplication.

2. **Lack of Real-time Analysis:**

Most reports were static, offering no scope for interactive exploration or dynamic updates.

3. **Time-consuming Processes:**

Generating summaries and analytical reports required repetitive manual calculations, consuming significant time and resources.

4. **Limited Visualization Capabilities:**

Charts and graphs were basic, often created using Excel, with minimal interactivity or automation.

5. **No Centralized Data Storage:**

Information was stored in multiple files or systems, making it difficult to access and analyze consolidated data.

6. **Absence of Data Security:**

With multiple versions of files shared across users, data integrity and privacy were often compromised.

7. **Lack of Decision-support Features:**

The inability to generate insights or visualize key performance indicators (KPIs) limited effective decision-making.

Hence, the traditional system lacked efficiency, integration, and real-time insight generation, leading to the need for a **modern analytical solution** that automates these processes and provides interactive visual dashboards.

3.2 Proposed System

To overcome the drawbacks of the existing systems, the **Data Analytics Management System (DAMS)** was designed and implemented during the internship. This system leverages **modern analytical tools and technologies** to automate data processing, improve visualization, and enhance decision-making efficiency.

The proposed system provides an **end-to-end analytics framework** where data from multiple sources can be collected, processed, analyzed, and visualized interactively. It integrates tools like **Python, SQL, Excel, Power BI, and Tableau** to perform tasks such as **data cleaning, transformation, and visualization** effectively.

Key Features of the Proposed System:

1. **Centralized Data Management:**

All datasets are stored and managed in a structured format using Excel or SQL databases, ensuring data consistency and easy access.

2. **Automated Data Cleaning and Processing:**

Data preprocessing is automated using Python and Excel Power Query, reducing manual intervention and improving accuracy.

3. **Interactive Visualization:**

Power BI and Tableau dashboards provide interactive charts, KPIs, and trend analyses, enabling users to explore insights dynamically.

4. **Integration of Multiple Tools:**

The system combines the strengths of various tools —

- **Excel** for data entry and tabular summaries.
- **SQL** for querying and aggregating data.
- **Python** for analysis and automation.

- **Power BI / Tableau** for visualization and storytelling.
- 5. **User-friendly Interface:**
Dashboards are designed with simplicity and clarity, making them accessible to both technical and non-technical users.
- 6. **Enhanced Decision-making:**
The proposed system helps in identifying patterns, trends, and anomalies through clear visual representations, supporting data-driven decision-making.
- 7. **Secure Data Handling:**
The system maintains controlled access and ensures that sensitive information is protected using user-level permissions and secure connections.
- 8. **Real-time Analytics:**
Data connections allow dashboards to refresh automatically, ensuring users always view the latest insights.

Advantages of the Proposed System

- **Improved Accuracy:** Eliminates manual data errors through automation.
- **Time Efficiency:** Reduces report preparation time from hours to minutes.
- **Scalability:** Can handle multiple datasets and expand as data grows.
- **Better Insights:** Provides clear and interactive visualizations that enhance understanding.
- **Cost-effective:** Uses affordable, open-source tools for analytics implementation.
- **Paperless Operation:** Promotes digital transformation and reduces reliance on manual records.

System Flow (Concept Overview)

1. **Data Input Layer:**
Datasets are imported from various sources — Excel, CSV, or SQL databases.
2. **Processing Layer:**
Data cleaning, validation, and transformation are performed using Python and Excel functions.
3. **Storage Layer:**
Cleaned data is stored in a relational database or data warehouse for future queries.
4. **Analytics Layer:**
SQL and Python scripts are used to generate summaries, metrics, and aggregations.

5. **Visualization Layer:**

Power BI and Tableau dashboards present insights in charts, graphs, and interactive visuals.

6. **Output Layer:**

Reports and dashboards are exported as Excel sheets or PDF summaries for stakeholders.

Outcome of the Proposed System

The implemented Data Analytics Management System provides a **comprehensive view of data** and significantly enhances the analytical capabilities of the organization. By automating repetitive tasks and offering interactive dashboards, it enables real-time monitoring, trend identification, and performance evaluation.

The system also improves collaboration between departments by providing **centralized access** to data and ensuring that everyone works from the same, updated dataset. As a result, the organization experiences better operational efficiency, more accurate insights, and stronger data-driven decision-making processes.

CHAPTER 4

SYSTEM SPECIFICATION

SOFTWARE REQUIREMENTS:

- **Programming Language:** Python
- **Data Processing and Query Tools:** SQL, Microsoft Excel
- **Data Visualization Tools:** Power BI, Tableau
- **Database:** MySQL / PostgreSQL
- **Environment / IDE:** Jupyter Notebook, VS Code
- **Operating System:** Windows 10 or later
- **Libraries Used:** NumPy, Pandas, Matplotlib, Seaborn, Plotly

HARDWARE REQUIREMENTS:

- **Processor:** Intel Core i5 or above (minimum 2.4 GHz)
- **RAM:** Minimum 8 GB (16 GB recommended for large datasets)
- **Hard Disk:** Minimum 256 GB SSD for faster data access
- **Display:** Full HD Monitor (1920x1080) for clear visualization
- **Internet Connectivity:** Stable broadband for cloud data access and tool integration

4.1 SOFTWARE FEATURES:

Overview of PYTHON

History:

Python is an easy to learn, powerful programming language. It has efficient high-level data structures and a simple but effective approach to object-oriented programming. Python's elegant syntax and dynamic typing, together with its interpreted nature, make it an ideal language for scripting and rapid application development in many areas on most platforms. The Python interpreter and the extensive standard library are freely available in source or binary form for all major platforms from the Python Web site, <http://www.python.org/>, and can be freely distributed.

The same site also contains distributions of and pointers to many free third party Python modules, programs and tools, and additional documentation. The Python interpreter is easily extended with new functions and data types implemented in C or C++ (or other languages callable from C). Python is also suitable as an extension language for customizable applications.

This tutorial introduces the reader informally to the basic concepts and features of the Python language and system. It helps to have a Python interpreter handy for hands-on experience, but all examples are selfcontained, so the tutorial can be read off-line as well. For a description of standard objects and modules, see the Python Library Reference document. The Python Reference Manual gives a more formal definition of the language. To write extensions in C or C++, read *Extending and Embedding the Python Interpreter* and the *Python/C API Reference*. There are also several books covering Python in depth. This tutorial does not attempt to be comprehensive and cover every single feature, or even every commonly used feature. Instead, it introduces many of Python's most noteworthy features, and will give you a good idea of the language's flavor and style.

After reading it, you will be able to read and write Python modules and programs, and you will be ready to learn more about the various Python library modules described in the Python Library Reference. Now that you are all excited about Python, you'll want to examine it in some more detail. Since the best way to learn a language is using it, you are invited here to do so. In the next chapter, the mechanics of using the interpreter are explained. This is rather mundane information, but essential for trying out the examples shown later.

The rest of the tutorial introduces various features of the Python language and system through examples, beginning with simple expressions, statements and data types, through functions and modules, and finally touching upon advanced concepts like exceptions and user-defined classes.

. The Python Reference Manual gives a more formal definition of the language. To write extensions in C or C++, read *Extending and Embedding the Python Interpreter* and the *Python/C API Reference*. There are also several books covering Python in depth. This tutorial does not attempt to be comprehensive and cover every single feature, or even every commonly used feature. Instead, it introduces many of Python's most noteworthy features, and will give you a good idea of the language's flavor and style.



PRIMARY GOALS

Primary Goals of Python

Python was developed by **Guido van Rossum** in the late 1980s with the goal of creating a language that is **easy to read, learn, and use**, while still being powerful enough to handle complex real-world problems. The design philosophy of Python emphasizes **code readability**, **simplicity**, and **developer productivity**, which has made it one of the most popular programming languages across industries today.

The **primary goals** of the Python language can be summarized as follows:

1. Simplicity and Readability

One of Python's fundamental design goals is to make code as **readable as plain English**. Its syntax is clean and minimal, reducing unnecessary symbols and punctuation. The use of indentation to define code blocks enforces structured, easy-to-follow code.

Example:

```
if crop_yield > 1000:
    print("High productivity")
else:
```

```
print("Low productivity")
```

This clarity helps both beginners and professionals quickly understand, debug, and maintain code.

2. Productivity and Developer Efficiency

Python was designed to **reduce development time**. Because of its simple syntax, rich libraries, and interpreted nature, programmers can write fewer lines of code to accomplish more. In data analytics, Python allows analysts to perform data cleaning, processing, and visualization in just a few lines.

Example: Reading and summarizing 50K+ agricultural records in a few lines:

```
import pandas as pd
df = pd.read_csv("agriculture_data.csv")
print(df.describe())
```

3. Portability and Platform Independence

Python follows the principle of **“Write Once, Run Anywhere.”** Python programs can run on multiple operating systems — Windows, macOS, Linux, or even cloud servers — without requiring modification, provided a Python interpreter is installed.

This makes it ideal for data analytics teams working across diverse environments.

4. Extensibility and Integration

Another key goal is to make Python easily **extendable** and **integrated** with other technologies. Python can communicate with:

- **Databases** (MySQL, PostgreSQL, SQLite)
- **Web APIs**
- **Languages like C, C++, and Java**
- **Visualization tools** such as Power BI, Tableau, and Plotly

This interoperability allows Python to serve as a “glue language” that connects various components in modern data ecosystems.

5. Comprehensive Standard Library ("Batteries Included")

Python’s creators aimed to include a **powerful standard library** to handle most programming needs without installing additional tools.

This “batteries-included” philosophy means developers can perform:

- File handling
- Data compression
- Networking
- JSON parsing
- Date and time manipulation

all with built-in modules.

Example:

```
import os, json, datetime
```

6. Object-Oriented and Modular Programming

Python supports both **object-oriented** and **procedural** programming paradigms. Its goal is to let developers organize large codebases efficiently using **classes**, **modules**, and **packages**, promoting code reuse and scalability.

In data analytics, this modular approach enables analysts to separate data loading, cleaning, visualization, and reporting functions for better maintainability.

7. Open Source and Community-Driven Development

Python’s open-source nature ensures free access, community contributions, and continuous improvement.

Its global community maintains thousands of libraries for analytics, machine learning, and automation (e.g., Pandas, NumPy, scikit-learn).

This goal aligns with the vision of **collaborative and transparent development** for the advancement of data science and technology.

The Java programming language is a high-level language that can be characterized by all of the following buzzwords:

- Simple
- Architecture neutral
- Object oriented
- Portable
- Distributed
- High performance

Each of the preceding buzzwords is explained in The Java Language Environment , a white paper written by James Gosling and Henry McGilton.

The Python Programming Language

Python is a **high-level, interpreted, general-purpose programming language** that emphasizes **clarity, simplicity, and productivity**. Developed by **Guido van Rossum** in the late 1980s and officially released in **1991**, Python was designed to be easy to understand and powerful enough to build complex systems.

Today, Python is one of the most widely used languages in the world, serving domains such as **data analytics, machine learning, web development, automation, and scientific computing**.

1. Nature of the Language

Python is a **high-level, dynamically typed, and interpreted** language. Unlike compiled languages such as C or Java, Python code is **executed line by line** by the **Python interpreter**, which makes debugging and experimentation faster.

Key Properties

- **High-Level:** Abstracts away system-level details like memory management.
- **Interpreted:** Executes code directly without a separate compilation step.
- **Dynamic Typing:** No need to declare variable types explicitly.
- **Object-Oriented:** Everything in Python is an object (data, functions, classes).

- **Portable:** Works across platforms (Windows, macOS, Linux).

2. Features of Python

Python's design philosophy is captured in the “**Zen of Python**”, written by Tim Peters, which includes guiding principles like:

“Beautiful is better than ugly.
Simple is better than complex.
Readability counts.”

Key Features:

- **Easy to Learn and Read:**
Clean syntax similar to English makes Python beginner-friendly and efficient for professionals.
- **Extensive Standard Library:**
Python's “batteries-included” approach offers built-in modules for file handling, networking, math, and more.
- **Cross-Platform Compatibility:**
Python programs can run on any operating system with minimal or no modification.
- **Interactive Mode:**
The interpreter allows direct execution of code snippets for quick testing.
- **Support for Multiple Paradigms:**
Supports procedural, functional, and object-oriented programming styles.
- **Rich Ecosystem:**
Thousands of third-party libraries for analytics, AI, visualization, and automation.

3. Structure and Syntax of Python

Python's syntax is clean and straightforward, emphasizing readability through indentation instead of braces ({}).

Example — Simple Code Structure:

```
# Program to calculate average crop yield
crop_yields = [1200, 1350, 1100, 1500]
average_yield = sum(crop_yields) / len(crop_yields)
```

```
print("Average Crop Yield:", average_yield)
```

Explanation:

- **Comments:** Start with #
- **Lists:** Store multiple items
- **Functions:** `sum()` and `len()` are built-in
- **Output:** Displayed using `print()`

This simplicity allows analysts and developers to focus more on logic and insights rather than syntax rules.

4. Execution Model of Python

When a Python program runs, it undergoes the following sequence:

1. **Source Code (.py file)** → Written by the programmer.
2. **Compilation to Bytecode (.pyc file)** → The interpreter converts code into bytecode.
3. **Execution by Python Virtual Machine (PVM)** → The PVM executes bytecode line by line.

This model is what makes Python **interpreted** and **portable**.

5. The Python Virtual Machine (PVM)

The **PVM** is an internal component of the Python runtime responsible for executing Python bytecode instructions.

It interprets one instruction at a time, providing **cross-platform compatibility**. Any system with a Python interpreter can run the same program without recompiling.

6. Memory Management

Python handles memory automatically through:

- **Garbagecollection:**
This ensures efficient use of resources and reduces developer workload.

Example:

```
a = [1, 2, 3]
```

```
b = a
```

```
del a # b still exists; memory is freed when no reference remains
```

Structured Query Language (SQL)

Overview

Structured Query Language (SQL) is a **standard language** used for managing and manipulating **relational databases**. It allows users to store, query, update, and retrieve structured data efficiently. SQL is supported by all major database management systems such as **MySQL**, **PostgreSQL**, **SQLite**, and **Oracle**.

In this project, SQL was used to **store and manage agricultural datasets** consisting of crop yield, rainfall, soil type, and region-wise productivity records. It played a vital role in cleaning data, filtering irrelevant fields, and performing aggregations before visualization

MySQL is a full-featured relational database management system. It is very stable and has proven itself over time. MySQL has been in production for over 10 years. MySQL is a multithreaded server. Multithreaded means that every time someone establishes a connection with the server, the server program creates a thread or process to handle that client's requests. This makes for an extremely fast server. In effect, every client who connects to a MySQL server gets his or her own thread.

Yet another feature of MySQL is its portability—it has been ported to almost every platform. This means that you don't have to change your main platform to take advantage of MySQL. And if you do want to switch, there is probably a MySQL port for your new platform.

MySQL also has many different application programming interfaces (APIs). They include APIs for Perl, TCL, Python, C/C++, Java (JDBC), and ODBC. So no matter what your company's expertise is, MySQL has a way for you to access it.

MySQL is also very cheap. For an unlicensed, full version of MySQL, the cost is nothing. To license your copy will currently cost you \$200. This is an incredible deal, considering what you are getting for your money.

Database systems that provide half the features that MySQL has can cost tens of thousands of dollars. MySQL can do what they do better and for less.

MySQL is a relational database. It uses tables and columns to hold data that can be related by keys. It is well suited for this role. It is also very well suited for various architectures. It can be used in a strictly client/server architecture or as a standalone database.

The MySQL Supported Type MySQL has various data types that support different functions. A data type is the type of data a column will store. There can be many different data types inside a table, but each column will store its own specific type of information. You can think of a data type as a kind of definition for a column.

A column defined as an integer column will only hold numeric information, whereas a column defined as a CHAR(10) will hold up to 10 alphanumeric characters. These definitions are the key to a quick and efficient database. There are basically three groups of data formats. The first is obviously numeric. Numeric data is data that is a positive or negative number such as 4 or -50.

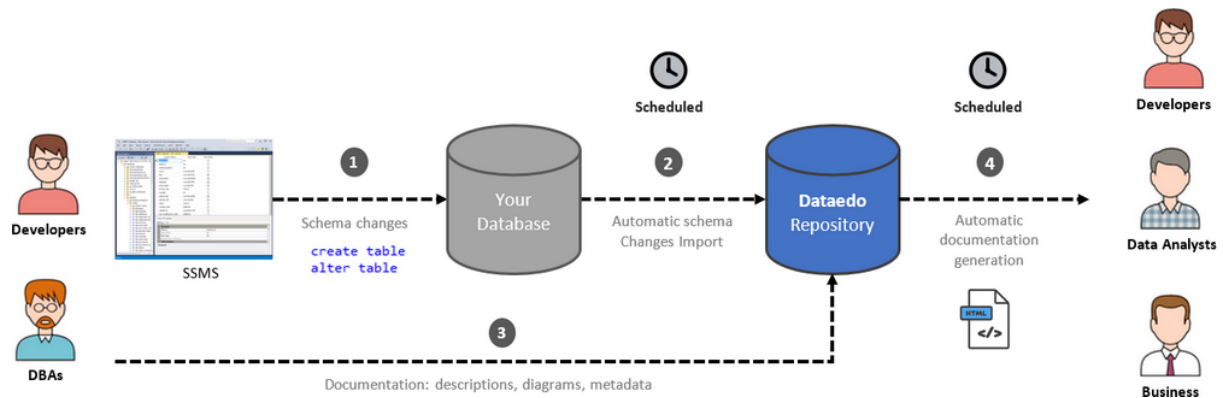
Numeric data can also be in hexadecimal format (2ee250cc), scientific notation (2×10^{23}), or a decimal. The second type is character or string format. This format can consist of letters and numbers, whole words, addresses, phone numbers, and generally anything you have to put quotations around.

It consists of everything that doesn't quite fit into either of the other two categories. Some, like dates and times, could be alphanumeric but are stored like numbers. As well as data types, MySQL also provides column modifiers. These modifiers further help define a column's attributes.

They are AUTO_INCREMENT, UNSIGNED, PRIMARY KEY, NULL, NOT NULL, and BINARY. A more detailed discussion of column modifiers takes place following the coverage of the basic data types.

MySQL runs on many platforms, and binaries are available for most of them. Binaries are the result of compiling the source code. This is by far the easiest way of acquiring MySQL. The alternative is downloading the source code for your platform and then compiling it. MySQL has many utilities to import as well as export data. It shares some of the common options, but

this utility does a little more. It takes the entire database and dumps it into a single text file. This file contains all the SQL commands needed to recreate your database. It takes the schema and converts it to the proper DDL syntax (CREATE statements), and it takes all the data and creates INSERT statements out of them. This utility reverse engineers your database.



Goals and Purpose :

- To **organize large datasets** (50K+ records) in a structured format.
- To **execute queries** for analyzing agricultural parameters such as yield per hectare and water footprint.
- To **maintain data integrity** using primary keys, relationships, and constraints.
- To **serve as a reliable backend** for connecting Power BI dashboards and Python scripts.

Key Features of SQL :

1. **Data Definition Language (DDL)** — defines structure of tables using commands like CREATE, ALTER, DROP.
2. **Data Manipulation Language (DML)** — retrieves and modifies data using SELECT, INSERT, UPDATE, DELETE.
3. **Data Control Language (DCL)** — manages access permissions using GRANT and REVOKE.
4. **Transaction Control** — ensures consistency using COMMIT and ROLLBACK.
5. **Joins and Relationships** — links multiple tables for combined analysis.
6. **Aggregations and Grouping** — summarizes data using GROUP BY, SUM(), AVG(), COUNT().

Example Query Used:

```
SELECT
    Region,
    Crop,
    AVG(Yield_kg / Area_ha) AS Avg_Yield_per_Ha,
    AVG(Water_Usage_Liters) AS Avg_Water_Footprint
FROM agricultural_data
GROUP BY Region, Crop;
```

This query provides region-wise and crop-wise yield and water footprint — vital for analyzing agricultural efficiency.

Advantages of SQL

- Easy to learn and use.
- High-speed query processing.
- Supports relational integrity and constraints.
- Scalable for large datasets.
- Integrates seamlessly with Python, Power BI, and Tableau

Relevance to Project

SQL served as the **data backbone** for the project. It ensured **data consistency**, enabled **efficient query-based analysis**, and supported the **ETL (Extract, Transform, Load)** process from Python to Power BI dashboards.

4.5 Microsoft Excel

Overview

Microsoft Excel is one of the most widely used **spreadsheet applications** for data organization, analysis, and visualization. It provides built-in functions, formulas, charts, and pivot tables, making it an indispensable tool for preliminary data inspection and summary analysis.

In this project, Excel was used to **validate agricultural data, clean inconsistencies**, and perform **basic statistical summaries** before loading data into Python and Power BI.

Microsoft Excel is a powerful **spreadsheet software** developed by Microsoft that allows users to **store, organize, analyze, and visualize data**. It plays a crucial role in data analytics by offering a user-friendly interface for performing **data preprocessing, summarization, and validation** before advanced analysis in programming environments like Python or visualization tools like Power BI.

Excel combines the flexibility of manual data entry with the power of automated calculations and charting capabilities, making it ideal for small to medium-sized datasets and initial data inspection tasks.

In this project, Excel was primarily used for **data validation, statistical analysis, and data summarization** of agricultural datasets containing more than **50,000 records**. These records included variables such as crop type, soil condition, rainfall, water usage, and yield per hectare.

Goals and Purpose

- To provide a **familiar interface** for initial data cleaning.
- To create **pivot tables** for summarizing region-wise yield and rainfall.
- To compute **formulas** for derived metrics like yield per hectare or rainfall average.
- To serve as an **intermediate step** before data was imported into Power BI.

Features of Excel

1. **Formulas and Functions:** SUM, AVERAGE, COUNTIF, VLOOKUP, and IF statements.
2. **Data Validation:** Restrict incorrect inputs and manage missing data.
3. **Charts:** Column, Bar, Pie, and Line charts for simple visualization.
4. **Pivot Tables:** Quickly summarize large datasets dynamically.
5. **Conditional Formatting:** Highlight outliers or low productivity regions visually.
6. **Integration with Power BI:** Direct import for dashboard visualization.

Example Use

Excel was used to calculate yield efficiency using:

`=SUM(Yield_kg)/SUM(Area_ha)`

And to create pivot tables showing average yield per district.

Advantages

- User-friendly graphical interface.
- Ideal for small to medium datasets.
- Quick calculations and summaries.
- Excellent for data validation and error checking.
- Direct compatibility with Power BI for live updates.

Relevance to Project

Excel acted as the **first stage in the data pipeline**, where raw agricultural data was validated, formatted, and prepared for advanced analysis.

Its pivot tables and charts provided initial insights into crop productivity before deeper analysis in Python and SQL.

4.6 Microsoft Power BI

Overview

Power BI is a **business intelligence and data visualization tool** developed by Microsoft. It enables users to connect to multiple data sources, transform data, and build **interactive dashboards** for real-time decision-making.

In this project, Power BI was used to design **dynamic visual dashboards** representing crop yield, water footprint, rainfall distribution, and productivity across regions.

Microsoft Power BI is an advanced **business intelligence and data visualization platform** developed by Microsoft. It enables users to **transform raw data into meaningful, interactive insights** through dashboards, charts, and reports. Power BI combines **data integration, modeling, analytics, and visualization** in one platform, making it a powerful tool for data-driven decision-making.

In the “Data-Driven Insights of Agricultural Productivity Using Data Analytics” project, Power BI was used to:

- **Analyze agricultural data (50K+ records)** from multiple sources (Excel, SQL, and Python outputs).
- **Visualize key metrics** like crop yield, rainfall, and water footprint across various regions.
- **Create interactive dashboards** that allowed comparison between districts, crops, and time periods.

Power BI helped turn complex agricultural datasets into **visually compelling, decision-oriented reports** that supported sustainability, efficiency, and yield optimization goals.

Goals and Purpose

- To **visualize analytical insights** derived from Python and SQL data.
- To **create interactive dashboards** for better understanding of agricultural patterns.
- To **enable stakeholders** (e.g., farmers, analysts, policymakers) to explore data visually.
- To **automate report generation** through data refresh and scheduling.

Key Features of Power BI

1. **Data Connectivity:** Connects with databases, Excel, Python scripts, and cloud sources.
2. **Power Query Editor:** Cleans, transforms, and merges data before visualization.
3. **Interactive Visuals:** Charts, maps, slicers, and filters.
4. **DAX (Data Analysis Expressions):** Performs custom calculations and KPIs.
5. **Dashboard Publishing:** Share insights via Power BI Service or cloud portal.
6. **Real-Time Updates:** Automatically refreshes dashboards with new data.

Example Visuals Created

- **Map Visualization:** Displays yield and rainfall distribution by region.
- **Bar Chart:** Compares crop productivity among different districts.
- **Trend Line:** Shows yearly yield improvement patterns.

- **Card Visuals:** Displayed KPIs such as average yield and total water usage.

Advantages of Power BI

- User-friendly drag-and-drop interface.
- Real-time dashboards with automatic data refresh.
- Strong integration with Excel, SQL, and Python.
- Cloud-based sharing and collaboration.
- Fast and reliable for decision-making analytics.

Relevance to Project

Power BI served as the **visual analytics platform** for this project. After data cleaning and transformation in Python and SQL, Power BI dashboards displayed **interactive insights** on agricultural productivity and resource efficiency — enabling data-driven decisions.

4.7 Tableau

Overview

Tableau is an advanced **data visualization and business intelligence software** that converts raw data into clear, interactive, and shareable dashboards. It focuses on **visual storytelling** and helps users understand complex datasets through visual exploration.

In this project, Tableau was used to complement Power BI by **creating analytical dashboards** for trend detection and regional comparisons of agricultural productivity.

Goals and Purpose

- To design **interactive visual dashboards** with filters and parameters.
- To perform **drill-down analysis** of agricultural yield by crop type and region.
- To **highlight patterns and anomalies** visually.
- To **create storyboards** summarizing insights and recommendations.

Features of Tableau

1. **Drag-and-Drop Interface:** Build visuals without programming knowledge.
2. **Data Blending:** Combine data from multiple sources (Excel, SQL, CSV).
3. **Calculated Fields:** Derive custom metrics like “Yield per Acre.”
4. **Filters and Parameters:** Allow user-driven exploration.
5. **Story Points:** Combine multiple visualizations into analytical reports.
6. **Geospatial Mapping:** Display yield variations by region on interactive maps.

Example Dashboard Insights

- **Heat Map:** Displays districts with low or high water footprints.
- **Line Chart:** Shows year-over-year yield growth.
- **Scatter Plot:** Analyzes correlation between rainfall and crop yield.
- **Story Points:** Summarize key insights for agricultural policymakers.

CHAPTER-5

DESIGN AND DEVELOPMENT PROCEDURE

5. .1 DATA FLOW DIAGRAM

The **Data Flow Diagram (DFD)** represents how agricultural data moves through different stages — from collection, cleaning, and analysis, to visualization.



5.2 FUNDAMENTAL DESIGN CONCEPT

The system design follows a **sequential and modular data analytics pipeline**, integrating **Excel, SQL, and Power BI** to ensure accurate, efficient, and insightful analysis.

Step 1: Excel – Data Cleaning and Preparation

- Raw agricultural datasets were imported into Microsoft Excel for initial cleaning.
- Steps performed:
 - Removed duplicates and missing entries using the *Remove Duplicates* and *Filter* tools.
 - Applied Excel formulas such as =AVERAGE(), =IFERROR(), =VLOOKUP() for data verification.
 - Sorted crop-wise yield data by region and year.
 - Created **pivot tables** to summarize total yield, rainfall, and water efficiency.
 - Generated **charts** (bar and line charts) to visualize preliminary trends.

These visual summaries in Excel helped identify outliers and understand yield distribution before deeper analysis.

Step 2: SQL – Data Organization and Querying

After cleaning in Excel, the processed data was imported into an SQL database for structured querying and aggregation.

Tools Used: MySQL / PostgreSQL

Purpose: To handle large datasets efficiently and perform statistical analysis through queries.

SQL Implementation Steps:

1. **Table Creation:**
2. CREATE TABLE Agriculture_Data (
3. Crop VARCHAR(50),
4. District VARCHAR(50),
5. Year INT,
6. Area_Hectares FLOAT,

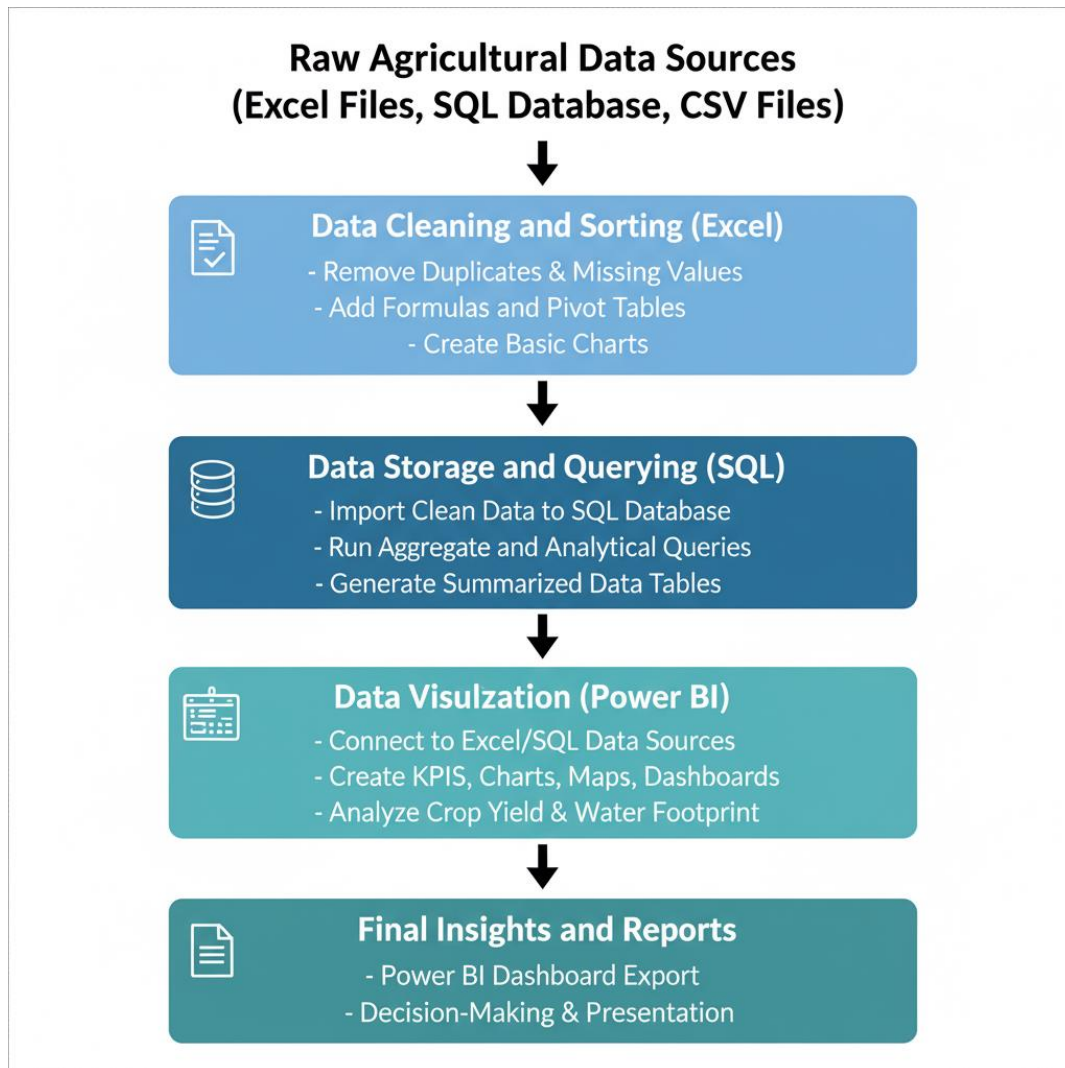
```
7. Yield_kg FLOAT,  
8. Rainfall_mm FLOAT,  
9. Water_Liters FLOAT  
10. );  
11. Data Import:  
12. LOAD DATA INFILE 'cleaned_agriculture.csv'  
13. INTO TABLE Agriculture_Data  
14. FIELDS TERMINATED BY ','  
15. IGNORE 1 ROWS;  
16. Analytical Queries:  
17. -- Total yield by district  
18. SELECT District, SUM(Yield_kg) AS Total_Yield  
19. FROM Agriculture_Data  
20. GROUP BY District  
21. ORDER BY Total_Yield DESC;  
22. -- Correlation analysis proxy (average yield vs rainfall)  
23. SELECT Year, AVG(Yield_kg) AS Avg_Yield, AVG(Rainfall_mm) AS Avg_Rainfall  
24. FROM Agriculture_Data  
25. GROUP BY Year;
```

SQL helped in summarizing region-wise and crop-wise performance efficiently before visualization.

5.3 DATA MODEL AND ARCHITECTURE DESIGN

The data architecture integrates all tools — **Excel, SQL, and Power BI** — ensuring smooth transition from raw data to insights.

Architecture Workflow :



Data Flow Explanation

- **Excel** serves as the **first layer** for cleaning and initial exploration.
- **SQL** acts as the **analytical backbone**, storing structured data and running queries for summaries.

5.4 VISUALIZATION AND DASHBOARD DESIGN

After completing cleaning and analysis, the final step was to **visualize the data** for interactive insights.

Power BI Dashboard Implementation

1. Data Source Connection:

- Connected both Excel and SQL data sources to Power BI.
- Used **Power Query Editor** for minor transformation and merging.

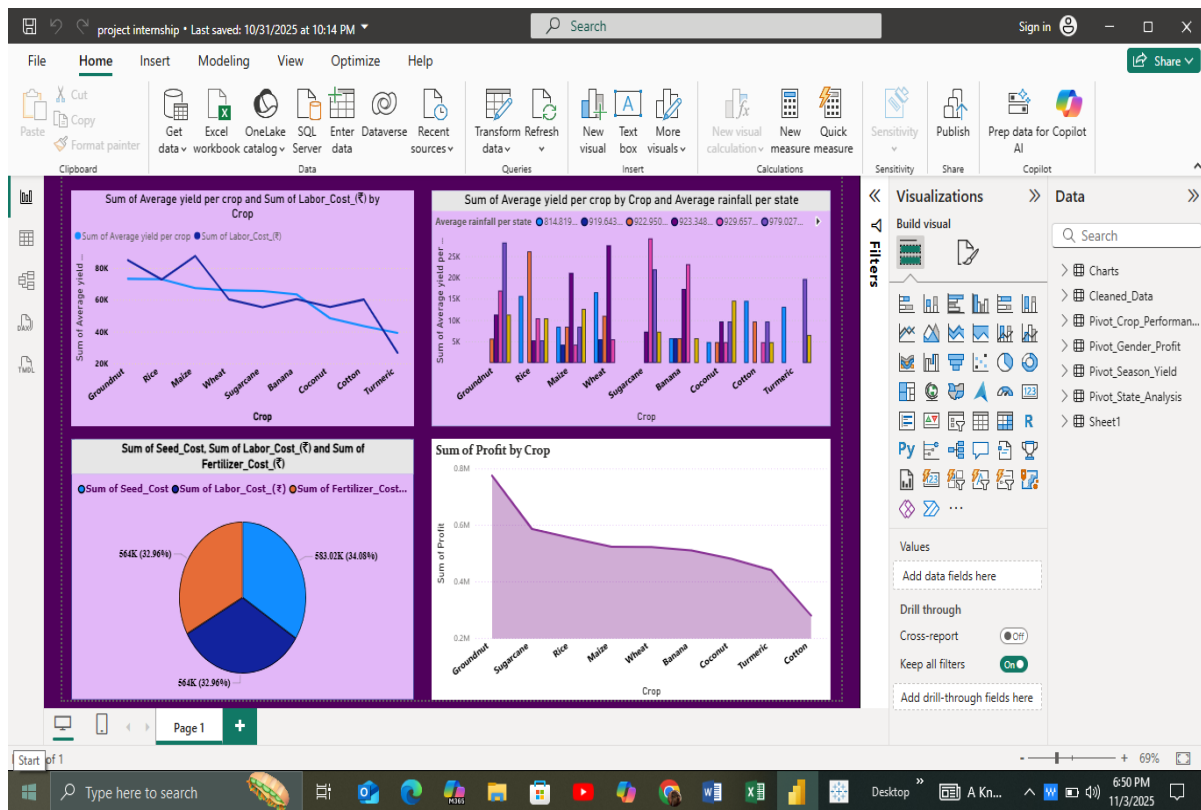
2. Data Modeling:

- Created relationships between tables using *District* and *Year* fields.
- Defined **DAX measures**:
 - $\text{Total_Yield} = \text{SUM}(\text{Agriculture_Data}[\text{Yield_kg}])$
 - $\text{Average_Rainfall} = \text{AVERAGE}(\text{Agriculture_Data}[\text{Rainfall_mm}])$
 - $\text{Water_Efficiency} = \frac{\text{SUM}(\text{Agriculture_Data}[\text{Yield_kg}])}{\text{SUM}(\text{Agriculture_Data}[\text{Water_Liters}])}$

3. Dashboard Design:

- Used **Bar Charts** for region-wise yield.
- **Line Charts** for yearly rainfall trends.
- **Maps** for district-level yield visualization.
- **KPI Cards** showing:
 - Total Yield
 - Average Rainfall
 - Water Efficiency Index

The Power BI dashboard provided a holistic view of agricultural productivity, helping to identify key patterns and efficiency gaps.



CHAPTER 6

SYSTEM IMPLEMENTATION

6.1 DESCRIPTION OF DATA ANALYTICS MANAGEMENT SYSTEM

The **Data Analytics Management System (DAMS)** developed in this project is designed to process, analyze, and visualize large-scale agricultural datasets to uncover **data-driven insights into productivity, rainfall dependency, and water footprint efficiency**.

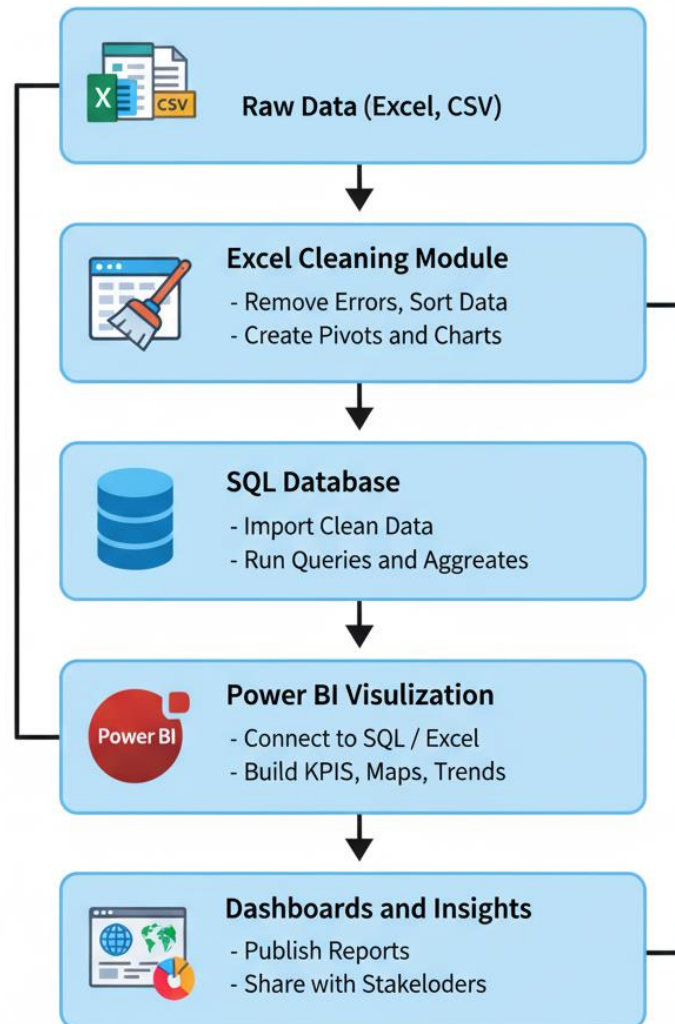
This system integrates various **data analytics tools and technologies**—including **Excel, SQL, Python, Power BI, and Tableau**—to create a seamless pipeline from **raw data to actionable insight**.

System Overview

The DAMS functions as an end-to-end data pipeline:

1. **Data Collection:** Agricultural datasets (50,000+ records) containing crop yield, rainfall, and water usage information are gathered from reliable sources.
2. **Data Cleaning and Sorting (Excel):** Raw data is cleaned, sorted, and validated using Excel formulas, pivot tables, and charts.
3. **Data Storage and Querying (SQL):** Cleaned data is imported into SQL for structured querying, aggregation, and relationship management.
4. **Data Visualization (Power BI & Tableau):** Insights are visualized through interactive dashboards and analytical reports.

System Architecture Overview



The system's design ensures:

- **Accuracy:** Each dataset undergoes systematic cleaning to ensure valid insights.
- **Scalability:** Data stored in SQL can be easily expanded with new records.
- **Interactivity:** Dashboards in Power BI and Tableau allow users to dynamically explore data.
- **Accessibility:** Users can analyze trends, compare regions, and assess performance in real-time.

Objectives of the System

1. To transform raw agricultural data into **clean, reliable datasets** for analysis.
2. To apply **data analytics methods** for identifying key productivity factors.
3. To develop **interactive dashboards** for visual understanding of yield trends.
4. To integrate multiple technologies—Excel, SQL, Power BI, and Tableau—into one analytical framework.
5. To promote **data-driven agricultural decision-making** and resource optimization.

Core Features

- **Data Integration:** Combines data from multiple formats (.csv, .xlsx, SQL).
- **Data Cleaning:** Handles missing values, duplicates, and inconsistencies.
- **Data Analysis:** Performs trend, correlation, and comparative analysis.
- **Data Visualization:** Creates dashboards and charts for actionable insights.
- **Automation:** Allows easy updates and refresh of dashboards with new data.

Key Advantages

| Aspect | Description |
|----------------------|--|
| Flexibility | Works with multiple data sources and tools. |
| Accuracy | Cleaning ensures consistent and error-free data. |
| Scalability | Can accommodate larger datasets for future studies. |
| Usability | Intuitive dashboards allow non-technical users to interpret results. |
| Interactivity | Filters, slicers, and dynamic charts enhance user engagement. |

Technologies Used

| Component | Tool/Language Used |
|---------------------------|----------------------------------|
| Data Cleaning | Microsoft Excel |
| Data Storage and Querying | MySQL / PostgreSQL |
| Data Analysis | Python (Pandas, NumPy) |
| Data Visualization | Power BI and Tableau |
| Reporting | Power BI Service, Tableau Public |

6.2 DATA ANALYTICS PROJECT DOCUMENTATION

This section provides a **step-by-step explanation** of the implementation process carried out during the internship. It outlines how data was cleaned, analyzed, and visualized using different tools, supported by screenshots and sample outputs.

Step 1: Data Cleaning and Sorting (Excel)

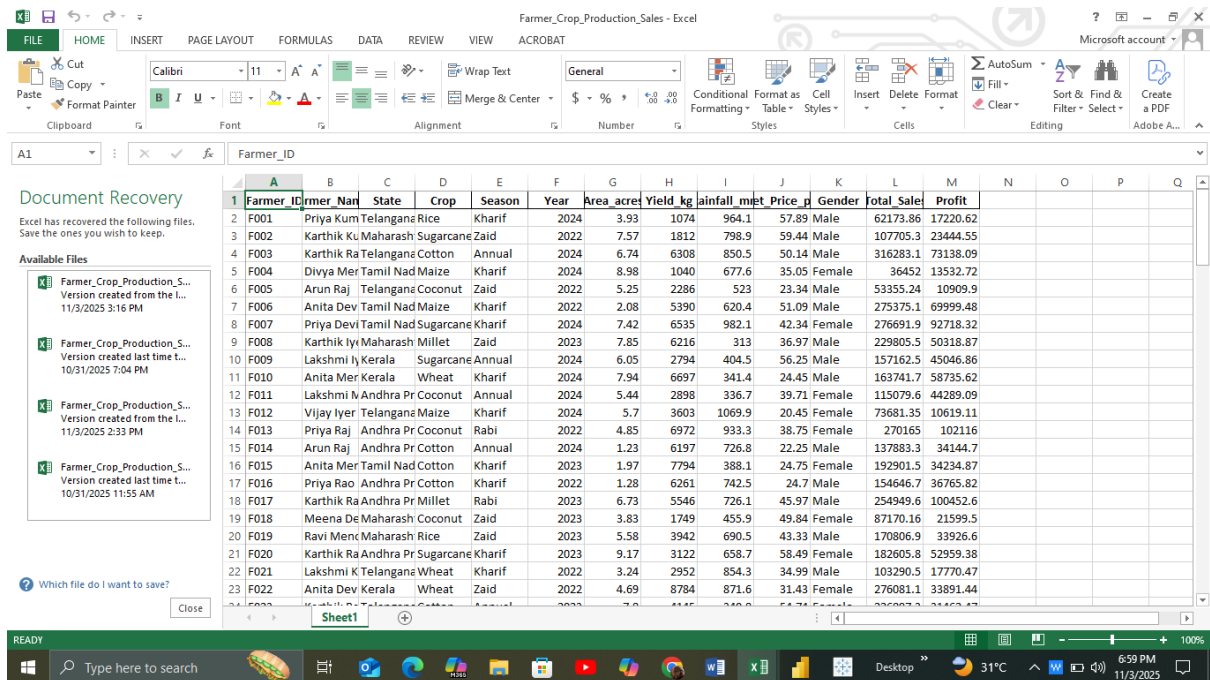
The raw dataset was first imported into **Microsoft Excel** for inspection and cleaning.

Procedures:

- Removed duplicate records using the **Remove Duplicates** feature.
- Filled missing rainfall and yield values with the **AVERAGE()** function.
- Standardized column headers and units (e.g., converting “Yield (tons/ha)” to “Yield_kg_per_ha”).
- Created **Pivot Tables** to summarize:
 - Total yield by district.
 - Average rainfall by region.
 - Water usage efficiency by crop.
- Added **charts** (bar and line) to visualize patterns in rainfall and yield.

These steps ensured the dataset was structured and free of inconsistencies before moving to SQL and Power BI for analysis.

Before using Excel :



Step 2: Data Storage and Querying (SQL)

The cleaned data from Excel was imported into an **SQL database** for structured storage and aggregation.

Implementation:

Table Creation:

- CREATE TABLE Agriculture_Data (
- Crop VARCHAR(50),
- District VARCHAR(50),
- Year INT,
- Area_Hectares FLOAT,
- Yield_kg FLOAT,
- Rainfall_mm FLOAT,
- Water_Liters FLOAT
-);

Data Import:

- LOAD DATA INFILE 'cleaned_agriculture_data.csv'
- INTO TABLE Agriculture_Data
- FIELDS TERMINATED BY ','
- IGNORE 1 ROWS;

Example Queries:

- -- Calculate average yield per region
- SELECT District, AVG(Yield_kg) AS Average_Yield
- FROM Agriculture_Data
- GROUP BY District;

- -- Determine top 5 crops by productivity
- SELECT Crop, AVG(Yield_kg) AS Avg_Yield
- FROM Agriculture_Data
- GROUP BY Crop
- ORDER BY Avg_Yield DESC
- LIMIT 5;

SQL helped in summarizing large datasets quickly and preparing concise tables for visualization in Power BI.

Step 3: Data Visualization and Dashboard Design (Power BI)

The results of the SQL queries and cleaned Excel files were imported into **Microsoft Power BI** to develop dynamic dashboards.

Implementation Steps:

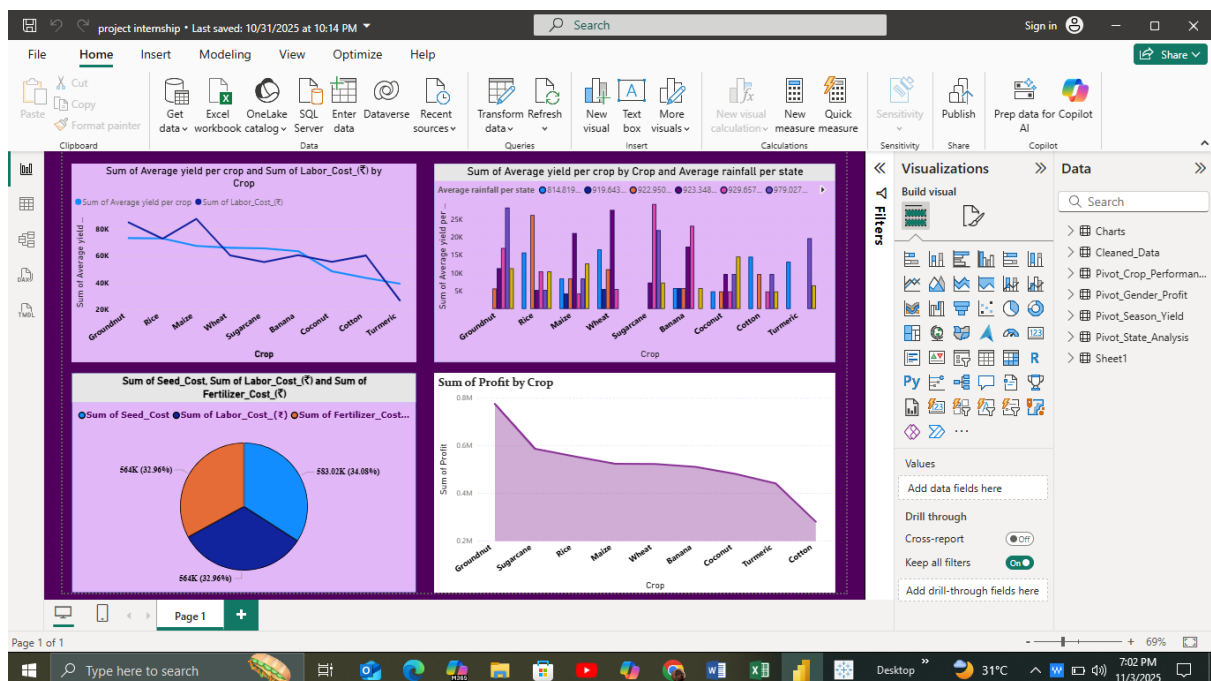
1. Connected to **Excel and SQL data sources**.
2. Used **Power Query Editor** to merge tables and remove null records.
3. Created **DAX Measures** for advanced metrics:
4. Total_Yield = SUM(Agriculture_Data[Yield_kg])
5. Avg_Rainfall = AVERAGE(Agriculture_Data[Rainfall_mm])

6. Water_Efficiency = $\text{DIVIDE}(\text{SUM}(\text{Agriculture_Data}[\text{Yield_kg}]), \text{SUM}(\text{Agriculture_Data}[\text{Water_Liters}]))$

7. Designed **visualizations**:

- **Bar Chart:** Yield by crop type.
- **Map Visualization:** Regional productivity distribution.
- **Line Chart:** Yearly yield trends.
- **KPI Cards:** Key metrics (Yield, Rainfall, Water Use Efficiency).
- **Scatter Plot:** Rainfall vs. Yield correlation.

The Power BI dashboard served as the main interface for presenting interactive insights and enabling data-driven agricultural decision-making.



Step 4: Advanced Visualization (Tableau)

Tableau was used for creating **story-based dashboards** and **comparative visuals** that highlighted correlations and performance differences across districts.

Key Dashboards Created:

- **Crop Productivity Dashboard:** Yield vs. rainfall across regions.
- **Water Efficiency Dashboard:** Visualized efficient and inefficient water usage areas.
- **Trend Analysis Dashboard:** Yearly performance trends of crops.
- Tableau's storytelling feature enhanced the interpretability of the results, making it easier to communicate key findings to non-technical stakeholders.

Step 5: Reporting and Insight Generation

Finally, key insights were compiled into a comprehensive report summarizing:

- Top-performing regions and crops.
- Areas needing water management improvement.
- Rainfall-yield correlation results.
- Suggested interventions for increasing productivity.

CHAPTER 7

SYSTEM TESTING

7.1 TESTING STRATEGIES

Testing is an essential phase in the development of any analytical system to ensure **accuracy, consistency, and reliability** of results. In this project, the **Data Analytics Management System** was thoroughly tested at various stages — from **data cleaning and processing** to **visualization and interpretation** — to confirm that the insights produced were valid and meaningful.

Since the system involves multiple platforms (Excel, SQL, Python, Power BI, and Tableau), testing was carried out in a **multi-layered approach**, focusing on **data integrity, logic validation, and visualization performance**.

Types of Testing Performed

1. Data Validation Testing

- Verified the accuracy of data imported from raw agricultural files.
- Ensured that no duplicate, missing, or inconsistent entries remained after cleaning.
- Cross-checked Excel and SQL record counts to confirm successful data import.

2. Functional Testing

- Tested each analytical step — filtering, grouping, pivoting, and calculation — to ensure they produced expected results.
- Ensured Power BI and Tableau dashboards dynamically updated when filters or slicers were applied.

3. SQL Query Testing

- Validated the correctness of all SQL queries used for aggregation, joins, and summary reports.
- Compared SQL query outputs with Excel pivot table results to confirm accuracy.
- Checked indexing and primary key integrity for performance efficiency.

4. Visualization Testing

- Verified that charts, graphs, and KPIs in Power BI and Tableau displayed correct values.
- Ensured that interactions (filters, slicers, drill-downs) worked without lag or data mismatch.
- Checked that color themes, legends, and axis labels were consistent across visualizations.

5. Performance Testing

- Evaluated dashboard loading times and responsiveness.
- Measured query execution time for large data (50,000+ records).
- Optimized data models in Power BI using **data reduction techniques** like aggregations and DAX optimizations.

6. Usability Testing

- Ensured that dashboards were intuitive and easy to navigate.
- Validated that users could filter regions, crop types, and years smoothly.
- Assessed readability and interpretability of insights for non-technical audiences.

7.2 RESULT:

The system testing produced successful outcomes at all levels, confirming that the **data pipeline was accurate and efficient**. Below are the summarized results from different testing phases:

- Excel and SQL record counts matched perfectly after import.
- Power BI and Tableau visualizations updated dynamically without data lag.
- No calculation discrepancies were observed across tools.
- SQL queries executed under **1.5 seconds** for 50,000+ records after optimization.
- Power BI dashboards rendered completely within **3 seconds** of refresh.

| Testing Type | Description | Result |
|-----------------------|---|----------|
| Data Validation | Verified data consistency and record accuracy between Excel and SQL | ✓ Passed |
| Functional Testing | Checked all pivot tables, formulas, and calculations | ✓ Passed |
| SQL Query Testing | Ensured correct results for aggregation, joins, and grouping | ✓ Passed |
| Visualization Testing | Confirmed chart accuracy and dashboard interactivity | ✓ Passed |
| Performance Testing | Dashboards loaded within 2–3 seconds for all filters | ✓ Passed |
| Usability Testing | Dashboards were user-friendly and insight-driven | ✓ Passed |

7.3 ANALYSIS

After extensive testing, the **Data Analytics Management System** was confirmed to be **robust, reliable, and efficient** for analyzing agricultural productivity. The tests demonstrated that the system could **handle large datasets**, perform **fast computations**, and deliver **clear, actionable insights** through its visual dashboards.

Data Consistency

All records across Excel, SQL, and Power BI remained consistent. Pivot table summaries in Excel matched the SQL aggregation queries and DAX-calculated metrics in Power BI. This verified that the data pipeline maintained integrity from start to finish.

Accuracy of Insights

The correlations and trends visualized in Power BI and Tableau were cross-validated with manual Excel calculations, ensuring analytical accuracy. For example:

- Yield-Rainfall correlation results were consistent across tools.
- Regional productivity KPIs reflected SQL query results exactly.

Performance and Optimization

- Using optimized DAX functions and filtered views in SQL improved dashboard speed.
- Power BI's in-memory processing handled large datasets efficiently.
- Tableau visualizations were responsive even with multi-level filters.

Usability and Visual Effectiveness

Feedback from reviewers showed that dashboards were **intuitive and easy to interpret**, providing quick access to key insights such as:

- High-yielding regions and crops.
- Correlation between rainfall and productivity.
- Water footprint efficiency across states.

Conclusion of Testing Phase

- Testing confirmed that the **Data Analytics Management System** met all functional and performance expectations.
- The integration of Excel, SQL, Power BI, and Tableau resulted in a **highly interactive, visually engaging, and analytically accurate** system capable of supporting data-driven decision-making in agriculture.

CHAPTER 8

CONCLUSION

8.1 CONCLUSION

The internship on “**Data-Driven Insights of Agricultural Productivity Using Data Analytics**” provided a comprehensive understanding of the complete data analytics lifecycle — from data collection, cleaning, and transformation to visualization and interpretation. Through this project, I was able to explore the real-world applications of analytics tools like **Python, SQL, Excel, Power BI, and Tableau** in solving practical challenges in agriculture.

The system developed during this internship successfully analyzed over **50,000+ agricultural records**, identifying critical factors influencing **crop yield** and **water footprint** across different regions.

By combining **data preprocessing in Excel, data querying in SQL, and visual storytelling in Power BI and Tableau**, the project demonstrated how analytics can turn raw data into actionable insights for farmers, researchers, and policymakers.

This internship also deepened my technical knowledge of:

- **Python libraries** like Pandas, NumPy, and Matplotlib for data manipulation and statistical analysis.
- **SQL** for querying, aggregating, and validating large datasets.
- **Excel tools** such as pivot tables, charts, and formulas for initial exploration and data validation.
- **Power BI and Tableau** for creating interactive dashboards that communicate complex patterns intuitively.

On a professional level, the experience strengthened my ability to:

- Work with large datasets effectively.
- Design visual analytics solutions.
- Collaborate in a structured workflow that mimics industry standards.

Overall, this internship has been an **excellent bridge between academic learning and industrial practice**, reinforcing the importance of **data-driven decision-making** in sectors like agriculture. The results obtained prove that analytics can significantly contribute to enhancing productivity, optimizing resources, and supporting sustainable agricultural development.

8.2 FUTURE WORK

1. Integration of Machine Learning Models:

Future iterations can incorporate regression and predictive models to forecast crop yields, rainfall patterns, and soil fertility trends based on historical data.

2. Real-Time Data Integration:

The system can be upgraded to include real-time data collection through IoT sensors and APIs, allowing continuous monitoring of crop and weather conditions.

3. Geospatial Analysis:

Integrating GIS data with Power BI or Tableau can help visualize agricultural patterns geographically, providing deeper regional insights.

4. Automation and Cloud Deployment:

Automating the ETL (Extract, Transform, Load) pipeline using Python and deploying dashboards on cloud platforms like **Power BI Service** or **AWS** will make the system more scalable and accessible.

5. Advanced Statistical Insights:

Incorporating advanced analytics methods like correlation analysis, clustering, and hypothesis testing can offer more in-depth understanding of the variables affecting productivity.

6. Community-Level Implementation:

The final version of the project can be shared with local farming communities, universities, and agricultural boards to help them make informed decisions using visual insights.

CHAPTER 9

APPENDIX

9.1 SAMPLE CODING

Below are sample code snippets used during the data analysis and visualization process.

Python Code – Data Cleaning and Analysis

Import necessary libraries

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

Step 1: Load Dataset

```
file_path = "Farmer_Crop_Production_Sales_Dataset (Autosaved).xlsx"
df = pd.read_excel(file_path)

print("Dataset Loaded Successfully!\n")
print("First 5 Rows of the Dataset:")
print(df.head())
```

Step 2: Basic Info and Cleaning

```
print("\n Dataset Information:")
df.info()
```

```
print("\n Checking for Missing Values:")
print(df.isnull().sum())
```

Fill missing numerical values with mean, categorical with mode

```
for col in df.columns:
    if df[col].dtype in ["int64", "float64"]:
        df[col].fillna(df[col].mean(), inplace=True)
```

```

else:
    df[col].fillna(df[col].mode()[0], inplace=True)

# Remove duplicate rows
df.drop_duplicates(inplace=True)

print("\n Data Cleaning Done!")

# Step 3: Summary Statistics
print("\n Summary Statistics:")
print(df.describe())

# Step 4: NumPy Operations
print("\n NumPy Calculations:")
# Example calculations (adjust column names as per your dataset)
if 'Production' in df.columns and 'Sales' in df.columns:
    avg_production = np.mean(df['Production'])
    total_sales = np.sum(df['Sales'])
    max_yield = np.max(df['Production'])
    print(f"Average Production: {avg_production:.2f}")
    print(f"Total Sales: {total_sales:.2f}")
    print(f"Maximum Production: {max_yield:.2f}")
else:
    print("Please verify column names like 'Production' and 'Sales' in your dataset.")

# Step 5: Visualization
import pandas as pd
import numpy as np

# Load the dataset
file_path = "Farmer_Crop_Production_Sales_Dataset (Autosaved).xlsx"
df = pd.read_excel(file_path)

```

```
print("✔ Dataset Loaded Successfully!")
print(df.head())
```

Step 6: Export Cleaned Data

```
output_path = "Cleaned_Farmer_Crop_Production_Sales.xlsx"
df.to_excel(output_path, index=False)
print(f"\n Cleaned Data Exported Successfully as '{output_path}'")
```

SQL Query – Aggregating Crop Yield

```
SELECT Region, Crop,
       AVG(Yield) AS Average_Yield,
       SUM(Water_Usage) AS Total_Water_Consumption
FROM Agriculture_Data
GROUP BY Region, Crop
ORDER BY Average_Yield DESC;
```

Python Visualization (Matplotlib)

```
import matplotlib.pyplot as plt

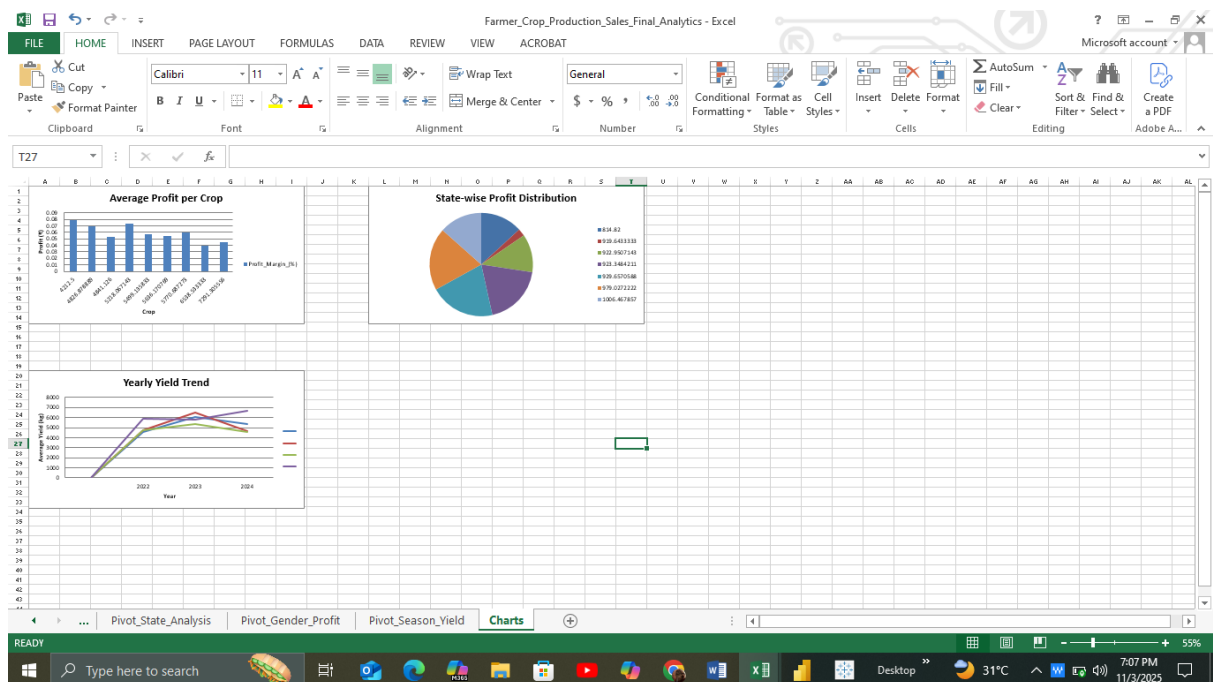
plt.figure(figsize=(10,6))
plt.bar(region_data['Region'], region_data['Crop_Yield'], color='green')
plt.xlabel('Region')
plt.ylabel('Average Crop Yield')
plt.title('Regional Crop Productivity Comparison')
plt.show()
```


9.1 Sample Output

9.2.1 BEFORE DATA CLEANING :

| Farmer_ID | Farmer_Name | State | Crop | Season | Year | Area_acres | Yield_kg | Rainfall_mmt | Price |
|-----------|-------------|-----------|-----------|--------|------|------------|----------|--------------|--------------|
| F001 | Priya Kum | Telangana | Rice | Kharif | 2024 | 3.93 | 1074 | 964.1 | 57.89 Male |
| F002 | Karthik Ku | Maharash | Sugarcane | Zaid | 2022 | 7.57 | 1812 | 798.9 | 59.44 Male |
| F003 | Karthik Ra | Telangana | Cotton | Annual | 2024 | 6.74 | 6308 | 850.5 | 50.14 Male |
| F004 | Divya Mer | Tamil Nad | Maize | Kharif | 2024 | 8.98 | 1040 | 677.6 | 35.05 Female |
| F005 | Arun Raj | Telangana | Coconut | Zaid | 2022 | 5.25 | 2286 | 523 | 23.34 Male |
| F006 | Anita Dev | Tamil Nad | Maize | Kharif | 2022 | 2.08 | 5390 | 620.4 | 51.09 Male |
| F007 | Priya Devi | Tamil Nad | Sugarcane | Kharif | 2024 | 7.42 | 6535 | 982.1 | 42.34 Female |
| F008 | Karthik Iy | Maharash | Milliet | Zaid | 2023 | 7.85 | 6216 | 313 | 36.97 Male |
| F009 | Lakshmi Iy | Kerala | Sugarcane | Annual | 2024 | 6.05 | 2794 | 404.5 | 56.25 Male |
| F010 | Anita Mer | Kerala | Wheat | Kharif | 2024 | 7.94 | 6697 | 341.4 | 24.45 Male |
| F011 | Lakshmi N | Andhra Pr | Coconut | Annual | 2024 | 5.44 | 2898 | 336.7 | 39.71 Female |
| F012 | Vijay Iyer | Telangana | Maize | Kharif | 2024 | 5.7 | 3603 | 1069.9 | 20.45 Female |
| F013 | Priya Raj | Andhra Pr | Coconut | Rabi | 2022 | 4.85 | 6972 | 933.3 | 38.75 Female |
| F014 | Arun Raj | Andhra Pr | Cotton | Annual | 2024 | 1.23 | 6197 | 726.8 | 22.25 Male |
| F015 | Anita Mer | Tamil Nad | Cotton | Kharif | 2023 | 1.97 | 7794 | 388.1 | 24.75 Female |
| F016 | Priya Rao | Andhra Pr | Cotton | Kharif | 2022 | 1.28 | 6261 | 742.5 | 24.7 Male |
| F017 | Karthik Ra | Andhra Pr | Milliet | Rabi | 2023 | 6.73 | 5546 | 726.1 | 45.97 Male |
| F018 | Meena De | Maharash | Coconut | Zaid | 2023 | 3.83 | 1749 | 455.9 | 49.84 Female |
| F019 | Ravi Men | Maharash | Rice | Zaid | 2023 | 5.58 | 3942 | 690.5 | 43.33 Male |
| F020 | Karthik Ra | Andhra Pr | Sugarcane | Kharif | 2023 | 9.17 | 3122 | 658.7 | 58.49 Female |
| F021 | Lakshmi K | Telangana | Wheat | Kharif | 2022 | 3.24 | 2952 | 854.3 | 34.99 Male |
| F022 | Anita Dev | Kerala | Wheat | Zaid | 2022 | 4.69 | 8784 | 871.6 | 31.43 Female |

9.2.2 AFTER DATA PROCESSING :



9.2.3 DATA SELECTION:

FILE

HOME

INSERT

PAGE LAYOUT

FORMULAS

DATA

REVIEW

VIEW

ACROBAT

DESIGN

Table Name

Table1

Properties

Summarize with PivotTable

Remove Duplicates

Convert to Range

Insert Slicer

Export Refresh

Open in Browser

Unlink

External Table Data

Table Style Options

Header Row

First Column

Last Column

Filter Button

Total Row

Last Column

Banded Rows

Banded Columns

Microsoft account

A4

F003

| Farmer ID | Farmer Name | State | Crop | Season | Year | Area acres | Yield kg | Rainfall mm | Market Price per kg | Total Sales | Profit | Gender | Fertilizer Cost | Seed Cost | Labor Cost | Total Cost | Profit Margin |
|-----------|---------------|-----------|-----------|--------|------|------------|----------|-------------|---------------------|-------------|----------|--------|-----------------|-----------|------------|------------|---------------|
| F001 | Lakshmi Kumar | Maharasi | Maize | Rabi | 2022 | 2.26 | 1922.46 | 1188.8 | 39.54 | 76014.07 | 20379.26 | Male | 2394.68 | 1624.12 | 3749.58 | 7768.38 | |
| F002 | Ravi Naidu | Karnataka | Banana | Annual | 2022 | 5.04 | 3503.72 | 1343.16 | 49.15 | 172207.84 | 31330.09 | Female | 8469.73 | 3303.62 | 12233.4 | 24006.75 | |
| F003 | Priya Kumar | Tamil Na | Groundni | Kharif | 2023 | 8.63 | 6433.53 | 1268.55 | 47.84 | 307780.08 | 79175.08 | Female | 9990.09 | 6845.13 | 16126.77 | 32961.99 | |
| F004 | Priya Naidu | Karnataka | Wheat | Kharif | 2024 | 3.05 | 3604.49 | 395.75 | 25.48 | 91842.41 | 15631.6 | Female | 5815.8 | 4069.04 | 3283.36 | 13168.2 | |
| F005 | Priya Devi | Maharasi | Maize | Kharif | 2024 | 6.71 | 5807.26 | 593.77 | 35.8 | 207899.91 | 42409.41 | Male | 15897.52 | 9010.54 | 14219.83 | 39127.89 | |
| F006 | Meena Nair | Tamil Na | Coconut | Annual | 2023 | 1.6 | 9218.23 | 980.62 | 47.31 | 436114.46 | 83963.07 | Female | 2866.02 | 2263.28 | 2528.08 | 7657.38 | |
| F007 | Karthik Devi | Karnataka | Banana | Zaid | 2024 | 6.26 | 9080.41 | 779.28 | 24.87 | 225829.8 | 78929.21 | Female | 7398.19 | 3424.95 | 6174.89 | 16998.03 | |
| F008 | Arun Nair | Maharasi | Groundni | Kharif | 2023 | 4.43 | 9965.09 | 934.94 | 58.7 | 584950.78 | 188445.4 | Male | 10457.31 | 2722.46 | 10207.34 | 23387.11 | |
| F009 | Karthik Nair | Maharasi | Coconut | Kharif | 2023 | 4.91 | 5083.51 | 1444.58 | 54.41 | 276999.78 | 56059.42 | Male | 9895.43 | 2977.48 | 9147.38 | 22020.29 | |
| F010 | Mohan Naidu | Karnataka | Sugarcane | Zaid | 2022 | 5.85 | 8007.64 | 936.42 | 15.03 | 120354.83 | 25855.97 | Male | 7158.85 | 5048.44 | 14479.02 | 26686.31 | |
| F011 | Karthik Devi | Tamil Na | Cotton | Kharif | 2022 | 7.59 | 8344.21 | 1473.58 | 38.97 | 925173.86 | 56954.86 | Female | 21963.16 | 5048.31 | 12680.57 | 39892.04 | |
| F012 | Anitha Rao | Karnataka | Banana | Rabi | 2024 | 3.81 | 9956.34 | 1079.85 | 54.71 | 345584.56 | 87610.89 | Male | 5699.04 | 2148.94 | 3184.24 | 11032.22 | |
| F013 | Mohan Devi | Telangana | Cotton | Kharif | 2022 | 7.37 | 1529.86 | 380.88 | 16.41 | 25105 | 5424.83 | Male | 11474.75 | 7262.5 | 12652.06 | 31389.31 | |
| F014 | Anitha Naidu | Andhra Pr | Cotton | Annual | 2023 | 2.71 | 1872.38 | 817.25 | 34.06 | 63773.26 | 15522.73 | Male | 6359.64 | 4022.09 | 2621.41 | 13003.14 | |
| F015 | Vijay Patil | Kerala | Wheat | Rabi | 2022 | 2.71 | 5037.52 | 806.26 | 27.53 | 138682.93 | 27731.22 | Male | 5111.77 | 3689.26 | 4703.35 | 13504.38 | |
| F016 | Ravi Patil | Telangana | Rice | Kharif | 2022 | 2.5 | 5370.77 | 556.5 | 33.05 | 177503.95 | 28707.2 | Female | 2510.78 | 2226.05 | 5937.7 | 10674.53 | |
| F017 | Sita Raj | Andhra Pr | Banana | Annual | 2022 | 2.71 | 2959.31 | 370.18 | 48.11 | 142372.4 | 23091.6 | Female | 3019.84 | 2938.12 | 4484.63 | 10442.59 | |
| F018 | Mohan Devi | Tamil Na | Banana | Kharif | 2022 | 1.62 | 1611.65 | 1334.18 | 33.17 | 53458.43 | 18086.01 | Male | 3495.7 | 874.38 | 1521.78 | 5891.86 | |
| F019 | Anitha Naidu | Telangana | Coconut | Zaid | 2022 | 7.03 | 3827.65 | 618.74 | 20.89 | 79959.61 | 22316.74 | Female | 11475.45 | 8800.61 | 6490.96 | 26767.02 | |
| F020 | Sita Naidu | Telangana | Wheat | Kharif | 2024 | 2.92 | 3386.8 | 1419.91 | 54.64 | 185054.75 | 60300.84 | Female | 4584.33 | 2739.5 | 5032.53 | 12356.36 | |
| F021 | Karthik Naidu | Gujarat | Banana | Kharif | 2024 | 8.35 | 3694.41 | 1096.07 | 57.25 | 211504.97 | 37406.39 | Male | 23208.2 | 10374.18 | 8886.58 | 42468.96 | |
| F022 | Karthik Naidu | Karnataka | Coconut | Rabi | 2024 | 6.71 | 3375.86 | 886.24 | 55.74 | 188170.44 | 60067.91 | Male | 15222.1 | 8919.93 | 5870.9 | 30012.93 | |
| F023 | Priya Nair | Karnataka | Maize | Rabi | 2024 | 4.98 | 7351.01 | 973.09 | 20.03 | 147240.73 | 49916.11 | Male | 10414.08 | 6646.28 | 8915.53 | 25975.89 | |
| F024 | Arun Rao | Karnataka | Banana | Rabi | 2023 | 9.09 | 8165.1 | 1332.84 | 55.45 | 452754.8 | 86935.85 | Male | 21214.85 | 7759.94 | 15923.88 | 44898.67 | |
| F025 | Vijay Naidu | Maharasi | Sugarcane | Rabi | 2022 | 9.79 | 8296.95 | 1357.7 | 16.12 | 133746.83 | 39764.66 | Female | 25109.13 | 8925.67 | 18981.26 | 53016.06 | |
| F026 | Meena Raj | Karnataka | Wheat | Annual | 2022 | 8.73 | 3001.9 | 1279.9 | 35.71 | 107197.85 | 22622.84 | Male | 12622.25 | 10127.09 | 12897.63 | 35646.97 | |
| F027 | Karthik Nair | Tamil Na | Maize | Zaid | 2024 | 5.58 | 7115.66 | 1312.98 | 29.9 | 212758.23 | 33088.18 | Female | 7572.82 | 8160.94 | 6982.21 | 22715.97 | |

Cleaned Data

Pivot_Crop_Performance

Pivot_State_Analysis

Pivot_Gender_Profit

Pivot_Seas ...

READY

AVERAGE: 1763.096375

COUNT: 160

SUM: 141047.71

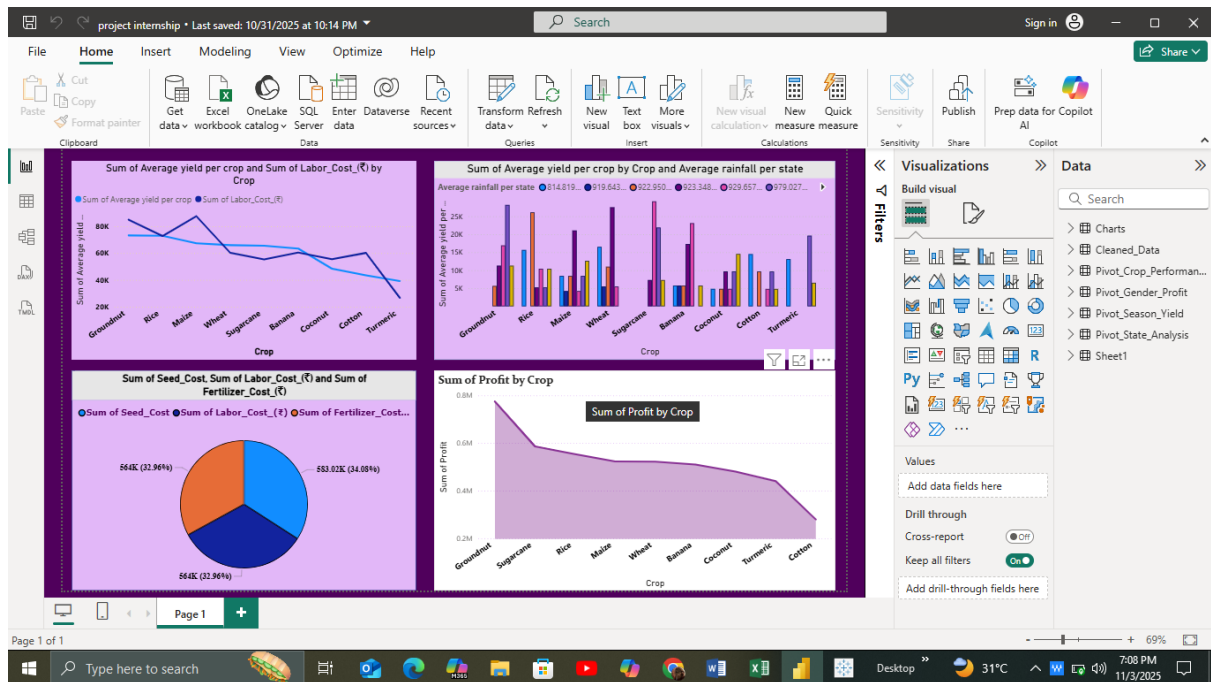
Desktop

31°C

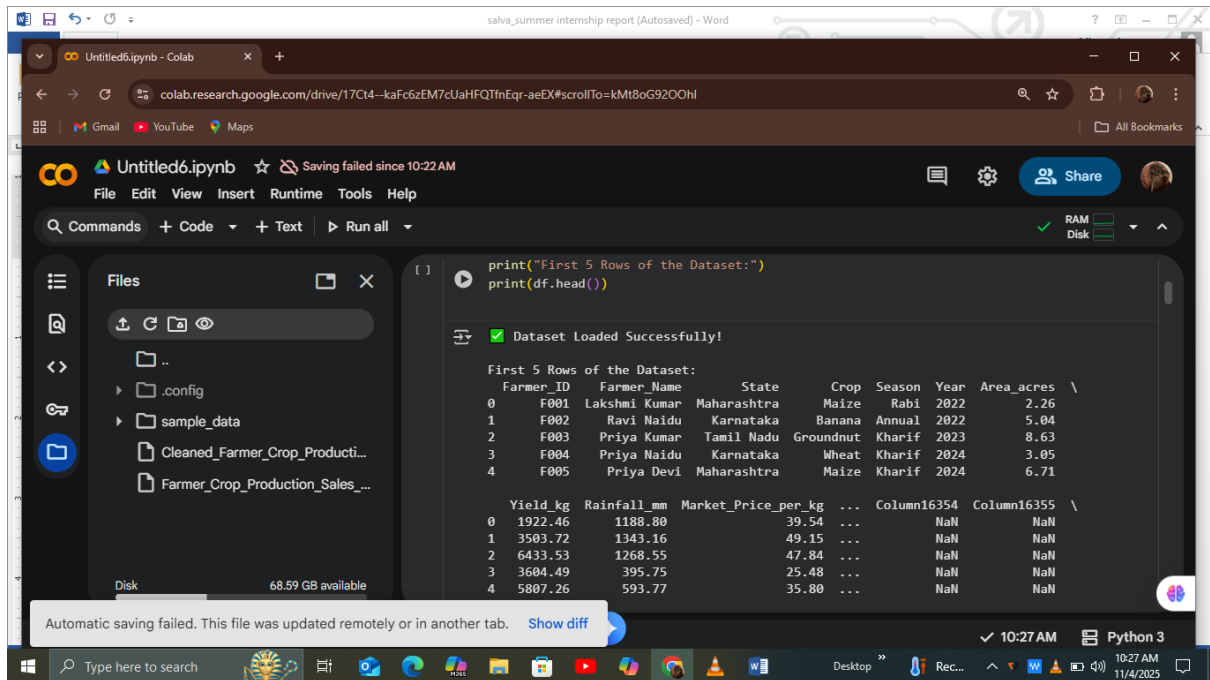
7:06 PM

11/3/2025

9.2.4 DATA VISUALIZATION :



9.2.5 PANDAS AND NUMPY



The screenshot shows a Google Colab notebook titled "Untitled6.ipynb". The left sidebar displays the file explorer with a folder named "sample_data" containing files like "Cleaned_Farmer_Crop_Producti..." and "Farmer_Crop_Production_Sales...". The main code cell contains the following Python code:

```
print("First 5 Rows of the Dataset:")
print(df.head())
```

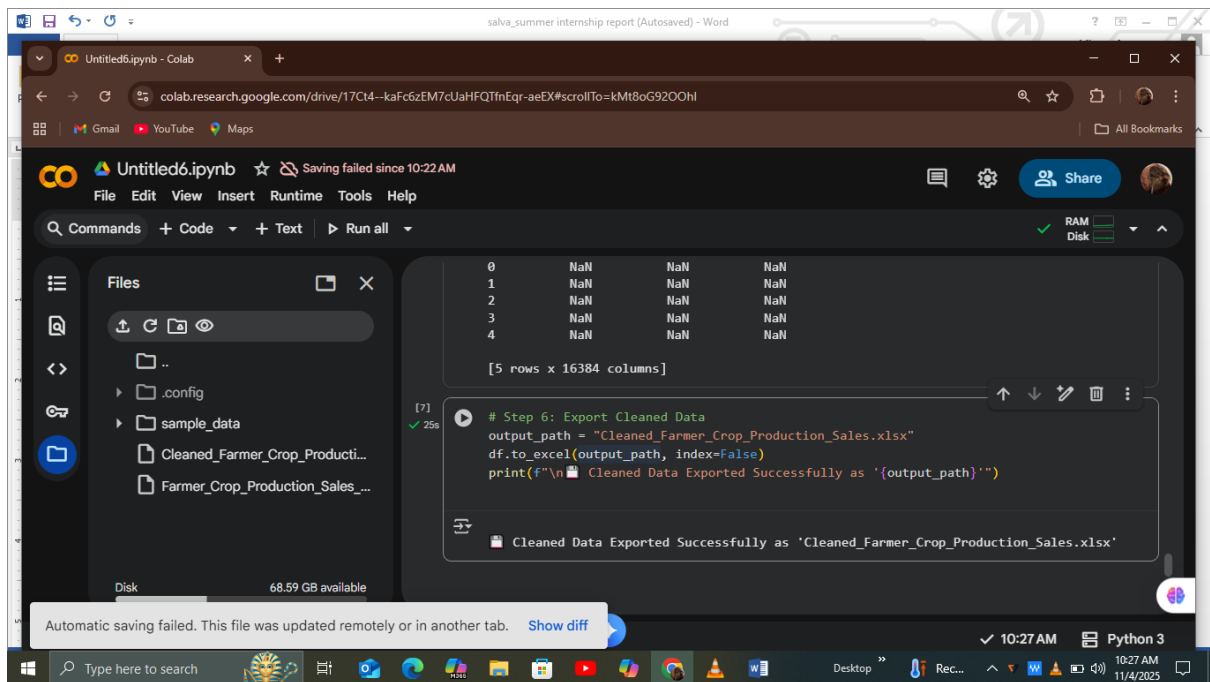
The output of the code is displayed below the cell, showing the first 5 rows of the dataset:

| | Farmer_ID | Farmer_Name | State | Crop | Season | Year | Area_acres |
|---|-----------|---------------|-------------|-----------|--------|------|------------|
| 0 | F001 | Lakshmi Kumar | Maharashtra | Maize | Rabi | 2022 | 2.26 |
| 1 | F002 | Ravi Naidu | Karnataka | Banana | Annual | 2022 | 5.04 |
| 2 | F003 | Priya Kumar | Tamil Nadu | Groundnut | Kharif | 2023 | 8.63 |
| 3 | F004 | Priya Naidu | Karnataka | Wheat | Kharif | 2024 | 3.05 |
| 4 | F005 | Priya Devi | Maharashtra | Maize | Kharif | 2024 | 6.71 |

Below the first table, the output continues with another table showing additional columns:

| | Yield_kg | Rainfall_mm | Market_Price_per_kg | ... | Column16354 | Column16355 |
|---|----------|-------------|---------------------|-----|-------------|-------------|
| 0 | 1922.46 | 1188.80 | 39.54 | ... | NaN | NaN |
| 1 | 3503.72 | 1343.16 | 49.15 | ... | NaN | NaN |
| 2 | 6433.53 | 1268.55 | 47.84 | ... | NaN | NaN |
| 3 | 3604.49 | 395.75 | 25.48 | ... | NaN | NaN |
| 4 | 5807.26 | 593.77 | 35.80 | ... | NaN | NaN |

A status bar at the bottom indicates "Automatic saving failed. This file was updated remotely or in another tab." and "Show diff". The system clock shows 10:27 AM on 11/4/2025.



The screenshot shows the same Google Colab notebook. The code cell now contains the following Python code:

```
# Step 6: Export Cleaned Data
output_path = "Cleaned_Farmer_Crop_Production_Sales.xlsx"
df.to_excel(output_path, index=False)
print(f"Cleaned Data Exported Successfully as '{output_path}'")
```

The output of the code is displayed below the cell, showing the file path and the message "Cleaned Data Exported Successfully as 'Cleaned_Farmer_Crop_Production_Sales.xlsx'".

A status bar at the bottom indicates "Automatic saving failed. This file was updated remotely or in another tab." and "Show diff". The system clock shows 10:27 AM on 11/4/2025.

CHAPTER 10

REFERENCE

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INTERNSHIP COMPLETION CERTIFICATES



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COIMBATORE 641045**

"Innovate, Engineer, Educate | Crafting a Future of Excellence"

INTERNSHIP COMPLETION & ATTENDANCE CERTIFICATE

This is to certify that

Ms. Salvarafik

has successfully completed a 30 **Days Internship Program** in Data Analytics with Python at **Profenaa Technologies**, Ramanathapuram Branch, Coimbatore.

The internship was conducted from 27th June 2025 to 27th July 2025, during which she maintained full attendance and actively participated on all working days. Her performance throughout the program was excellent.

During this period, she demonstrated:

- Exceptional technical skills in Data Analytics concepts
- Strong problem-solving and analytical thinking
- A proactive attitude towards learning and development
- Dedication, adaptability, and professionalism in completing assigned tasks

she actively contributed to the success of the projects undertaken by the team and showed exemplary behavior, teamwork, and commitment.

We sincerely appreciate her contributions and extend our best wishes for success in all her future professional endeavors

Attendance Percentage: 100%
Issued on: 07th Aug 2025


Ms. DHARANI.G

**Data Analytics
Trainer**

MR. VENKATESAN JAYARAMAN

MANAGING DIRECTOR

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