

Samsung Innovation Campus GSSSIETW, Mysore

Project Title: Employee Salary Data Cleaning and Aggregation for HR Insights

Description: The project “Employee Salary Data Cleaning and Aggregation for HR Insights” focuses on converting raw salary data into reliable and structured information for HR use. It involves cleaning missing and inconsistent values, standardizing records, and aggregating data such as average salary and departmental distribution. Visualizations like charts and graphs are used to highlight trends and patterns. This project supports HR teams in making data-driven decisions for workforce and compensation planning.

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Table of the content:

- Problem statement
- Objectives and features
- UML
- Detailed explanation
- Code explanation
- Output and explanation
- Usage instruction
- Troubleshooting
- Bibliography

Problem Statement: The HR department of a company maintains employee records, including salaries, departments, and job titles. However, the dataset contains duplicates, missing values, and inconsistent data formats.

The goal is to clean the employee dataset using Pandas, perform aggregation to extract useful HR insights, and export the results for reporting

Project Objectives

- **Data Cleaning:** Automated removal of duplicates and standardization of employee records
- **Data Enrichment:** Calculate derived metrics like years of service
- **Analysis:** Generate department and job title performance summaries
- **Visualization:** Interactive dashboard for HR insights and reporting

- **Export Capabilities:** CSV export functionality for further analysis

Features :

Key Performance Indicators

- Total Employees: Current employee count with filtering
- Average Salary: Mean salary across selected filters
- Total Salary Cost: Sum of all salaries
- Average Years of Service: Mean employee tenure

Interactive Filters

- Department Filter: Dropdown to select specific department or view all
- Real-time Updates: All metrics and charts update automatically

Visualizations

1. Salary Distribution Histogram

- Shows salary range distribution
- Identifies salary clusters and outliers
- Updates based on department filter

2. Department Salary Comparison

- Bar chart comparing average salaries by department
- Helps identify compensation disparities
- Visual department performance comparison

3. Salary vs Experience Scatter Plot

- Relationship between tenure and compensation
- Color-coded by department
- Identifies salary progression patterns

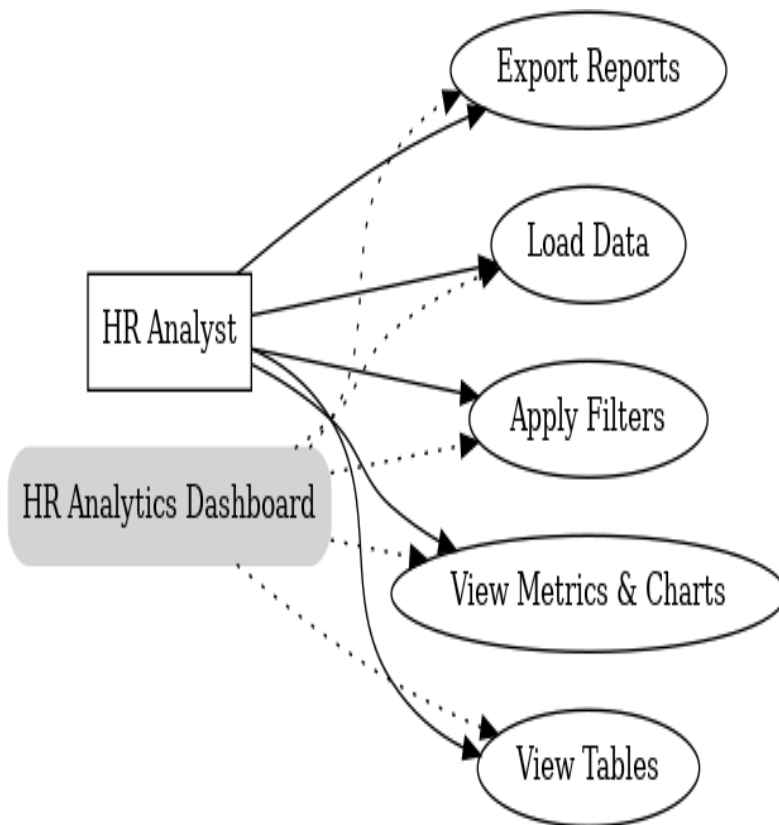
❑ Export Functionality

4. CSV Downloads: Export any table or filtered dataset
5. Custom File Names: Descriptive export file naming
6. Real-time Data: Always exports current filter state

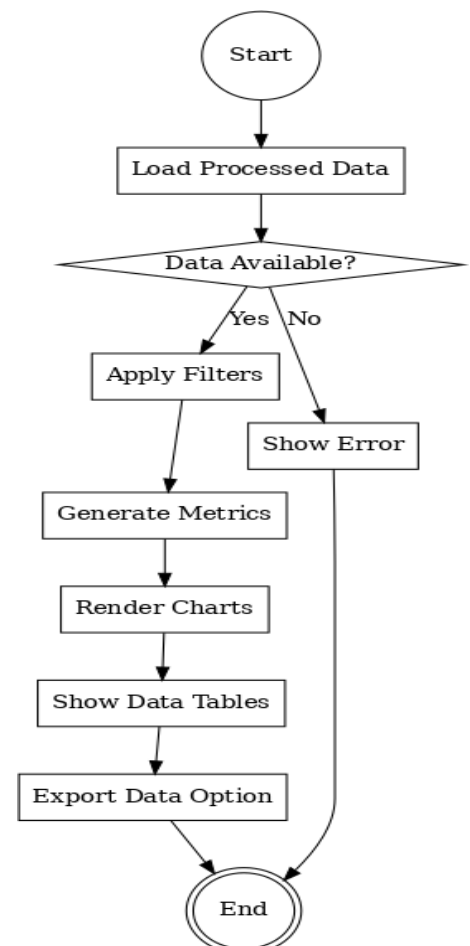
❑ Export Functionality

- CSV Downloads: Export any table or filtered dataset
- Custom File Names: Descriptive export file naming
- Real-time Data: Always exports current filter state

UML Diagram:



Use case diagram



Activity diagram

Structure of the project:

```
SIC_EmployeeInsights/
|
|— flask_app/
|   |— app.py           # Main Flask application
|   |
|   |— static/          # CSS, JS, images
|   |   |— css/
|   |   |   |— style.css
|   |   |— js/
|   |   |   |— script.js
|   |   |— images/
|   |
|   |— templates/       # HTML templates
|   |   |— index.html
|   |   |— dashboard.html
|   |   |— error.html
|   |
|   |— data/            # Dataset folder
|   |   |— employees.csv
|   |
|   |— utils/           # Helper functions
|   |   |— data_cleaning.py
|   |   |— data_analysis.py
|   |
|   |— __init__.py      # Makes flask_app a package
|
|— requirements.txt     # Python dependencies
|— README.md           # Project documentation
|— venv/               # Virtual environment (optional, not uploaded to GitHub)
```

Detailed Explanation:

□ Technical Stack

Core Technologies

- **Python 3.7+:** Main programming language
- **Pandas:** Data manipulation and analysis
- **Streamlit:** Interactive web dashboard framework
- **Matplotlib:** Data visualization
- **NumPy:** Numerical computing

Development Environment

- **VS Code:** Primary IDE
 - **Command Line:** Script execution and package management
 - **Git:** Version control
-

□ Installation & Setup

Prerequisites

- Python 3.13
- pip (Python package installer)

Step 1: Install Required Packages

`pip install pandas streamlit matplotlib numpy seaborn`

Step 2: Project Setup

1. Create project directory: `D:\learning\gsss_sic\mini_project`

2. Place your employees.csv file in the project directory
3. Copy all Python scripts to the project directory
4. Input Data Format

The system expects a CSV file named `employees.csv` with the following structure:

Column	Data Type	Description	Example
EmpID	String	Unique employee identifier	E001, E002
Name	String	Employee full name	Alice Wong
Department	String	Department name	HR, IT, Finance, Sales
JobTitle	String	Employee job title	HR Manager, Software Engineer
Salary	Integer	Annual salary in dollars	60000, 75000
JoiningDate	Date	Employee joining date	2019-03-15

Code Explanation:

Step 1: Data Loading & Inspection

File: `step1_load_inspect.py`

Purpose: Initial data exploration and quality assessment

Operations:

- Load CSV data into pandas DataFrame
- Display dataset information (`info()`, `head()`, `describe()`)
- Identify data types, missing values, and basic statistics

Output: Console display of dataset overview

Step 2: Data Cleaning & Standardization

File: step2_clean_standardize.py

Purpose: Clean and standardize raw data

Operations:

- Remove exact duplicate rows
- Handle duplicate Employee IDs (keep first occurrence)
- Fill missing salary values with median
- Replace missing job titles with "Unknown"
- Standardize department names (HR, IT, Finance, Sales)
- Standardize job title variations

Output: cleaned_employees.csv

Step 3: Data Enrichment

File: step3_add_derived_column.py

Purpose: Add calculated columns for enhanced analysis

Operations:

- Convert JoiningDate to datetime format
- Calculate Years of Service based on current date
- Handle date parsing errors gracefully

Output: employees_enriched.csv

Step 4: Department Analysis

File: step4_group_by_department.py

Purpose: Generate department-wise performance metrics

Operations:

- Group employees by department
- Calculate average salary per department
- Calculate total salary expenditure per department
- Count employees per department

Output: dept_summary.csv

Step 5: Top Performers by Department

File: step5_highest_paid_per_department.py

Purpose: Identify highest-paid employee in each department

Operations:

- Find maximum salary employee per department
- Extract detailed employee information
- Sort results by department

Output: highest_paid.csv

Step 6: Job Title Analysis

File: step6_group_by_jobtitle.py

Purpose: Analyze salary patterns by job title

Operations:

- Group employees by job title
- Calculate average salary per job title
- Calculate total salary cost per job title
- Count employees per job title

Output: job_summary.csv

Step 7: Automated Pipeline Execution

File: step7_all_exports.py

Purpose: Execute entire data processing pipeline

Operations:

- Run steps 2-6 sequentially
- Handle execution errors
- Provide pipeline status updates

Output: All CSV files generated automatically

Step 8: Dashboard Visualization

File: dashboard.py

Purpose: Provide an interactive dashboard for HR to view aggregated salary insights and anomalies

Operations:

- Load processed datasets (payroll_enriched.csv, dept_summary.csv, anomalies.csv)
- Display KPI metrics (total payroll, avg salary, median salary, headcount)
- Visualize charts:
 - Salary distribution (histogram)
 - Average salary by department (bar chart)
 - Headcount by department (bar chart)
- Allow export of filtered views (CSV)

Output:

Interactive dashboard tables, and downloadable reports for HR managers

Code all_exports.py:

```
step7_all_exports.py > ...
1 # Runs steps 2 -> 6 in one go. Assumes employees.csv is present.
2 import subprocess, sys
3 steps = [
4     "step2_clean_standardize.py",
5     "step3_add_derived_column.py",
6     "step4_group_by_department.py",
7     "step5_highest_paid_per_department.py",
8     "step6_group_by_jobtitle.py",
9 ]
10 for s in steps:
11     print(f"\n=== Running {s} ===")
12     ret = subprocess.call([sys.executable, s])
13     if ret != 0:
14         print(f"{s} failed with code {ret}")
15         sys.exit(ret)
16 print("\n✅ All outputs ready: cleaned_employees.csv, employees_enriched.csv, dept_summary.csv, highest_paid.csv, job_summary.csv")
17
```

Output and Explanation:

Output of step7 all_exports.py:

Explanation:

Step 2 cleans and standardizes raw employee data into a consistent format with basic employee information like ID, name, department, job title, salary, and joining date.

Step 3 adds a calculated "Years of Service" column to the cleaned data, showing how long each employee has been with the company.

Step 4 groups employees by department and creates summary statistics showing average salary, total salary costs, and employee count for each department.

Step 5 identifies the highest-paid employee in each department, revealing top earners like Marcus Allen (IT Data Scientist) at \$99,000.

Step 6 analyzes data by job title across all departments, showing salary ranges and employee counts for each position type.

The Finance department has the highest average salary, while IT has the most employees with 15 staff members.

Data Scientists earn the highest average salary at \$97,000, while Sales Executives are the most common job title with 8 employees.

The pipeline generates 5 CSV files providing different analytical views of employee data for HR decision-making and compensation analysis.

```
D:\learning\mini_project>python all_exports.py
python: can't open file 'D:\\learning\\mini_project\\all_exports.py': [Errno 2] No such file or directory

D:\learning\mini_project>python step7_all_exports.py

=== Running step2_clean_standardize.py ===
✓ Cleaned & standardized -> cleaned_employees.csv
EmpID Name Department JobTitle Salary JoiningDate
0 E001 Alice Wong HR HR Manager 60000 2019-03-15
1 E002 Bob Smith IT Software Engineer 75000 2020-06-20
2 E003 Carol Lee Finance Accountant 55000 2018-11-12
3 E004 David Kim IT Data Scientist 95000 2021-02-10
4 E005 Eva Brown Sales Sales Executive 50000 2019-07-01

=== Running step3_add_derived_column.py ===
✓ Added YearsOfService -> employees_enriched.csv
EmpID Name Department JobTitle JoiningDate YearsOfService
0 E001 Alice Wong HR HR Manager 2019-03-15 6
1 E002 Bob Smith IT Software Engineer 2020-06-20 5
2 E003 Carol Lee Finance Accountant 2018-11-12 6
3 E004 David Kim IT Data Scientist 2021-02-10 4
4 E005 Eva Brown Sales Sales Executive 2019-07-01 6

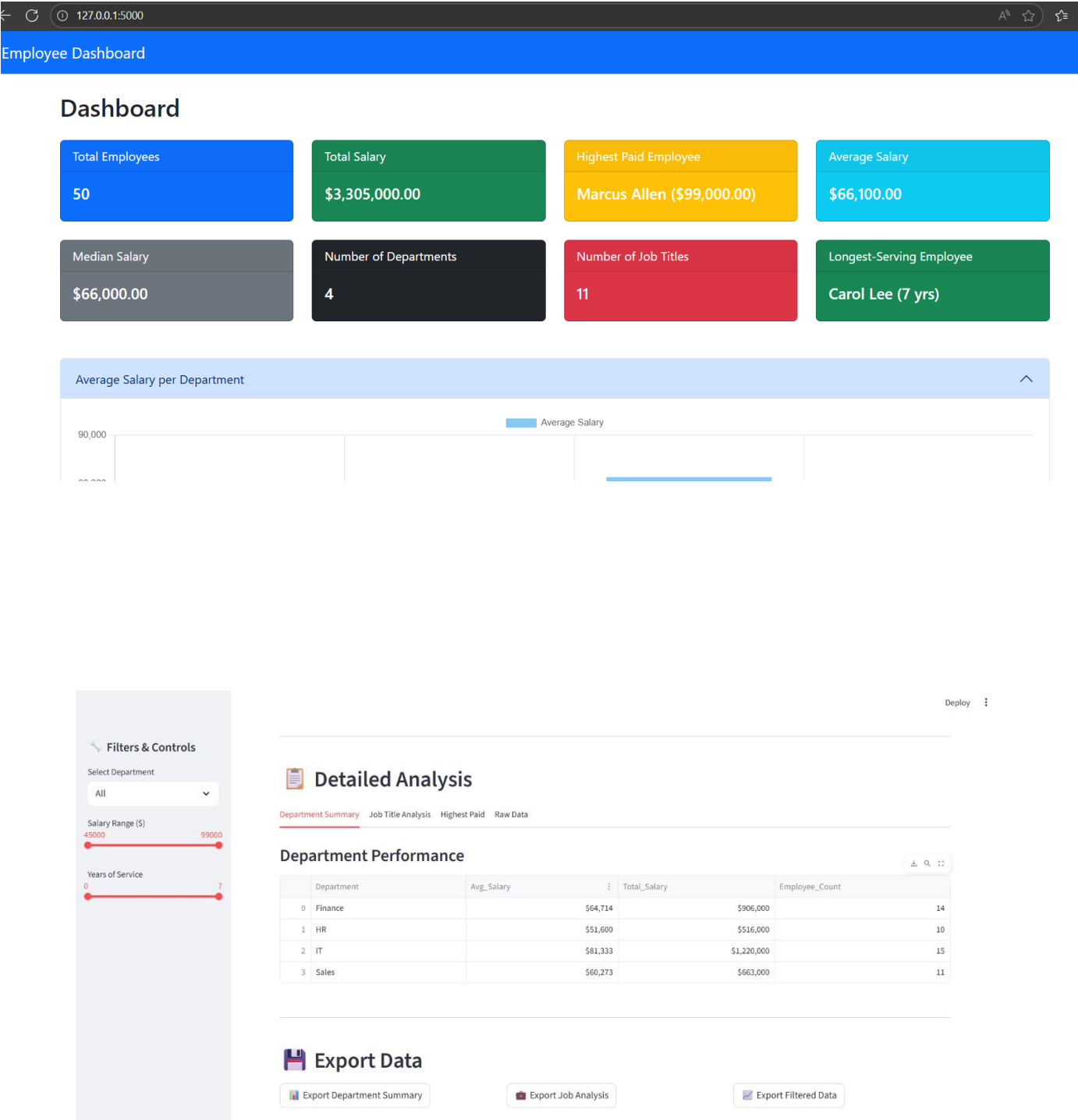
=== Running step4_group_by_department.py ===
✓ Department summary -> dept_summary.csv
Department Avg_Salary Total_Salary Employee_Count
0 Finance 64714.285714 906000 14
1 HR 51600.000000 516000 10
2 IT 81333.333333 1220000 15
3 Sales 60272.727273 663000 11

=== Running step5_highest_paid_per_department.py ===
✓ Highest paid per dept -> highest_paid.csv
Department EmpID Name JobTitle Salary
49 Finance E050 Yusuf Khan Senior Accountant 74000
37 HR E038 Lily Perez HR Manager 62000
38 IT E039 Marcus Allen Data Scientist 99000
25 Sales E026 Zach Miller Sales Manager 83000

=== Running step6_group_by_jobtitle.py ===
✓ Job title summary -> job_summary.csv
JobTitle Avg_Salary Total_Salary Employee_Count
1 Data Scientist 97000.000000 485000 5
7 Sales Manager 82000.000000 246000 3
9 Software Engineer 75428.571429 528000 7
8 Senior Accountant 72500.000000 290000 4
10 System Administrator 69000.000000 207000 3
2 Financial Analyst 66200.000000 331000 5
4 HR Manager 61000.000000 183000 3
0 Accountant 57000.000000 285000 5
6 Sales Executive 52125.000000 417000 8
3 HR Executive 49000.000000 147000 3
5 Recruiter 46500.000000 186000 4

✓ All outputs ready: cleaned_employees.csv, employees_enriched.csv, dept_summary.csv, highest_paid.csv, job_summary.csv
```

Output of dashboard.py:



Explanation:

This HR Analytics Dashboard represents the visual output of the employee data processing pipeline, transforming the raw CSV files into an interactive business intelligence interface.

Dashboard Overview The system displays comprehensive HR metrics with a total of 50 employees, an average salary of \$66,100, total salary costs of \$3.305 million, and an average tenure of 5.1 years across the organization.

Interactive Features The left sidebar provides filtering capabilities allowing users to select specific departments and adjust salary ranges (\$45,000-\$99,000) and years of service (1-7 years) to customize their analysis views.

Key Visualizations The first screen shows three primary charts: a salary distribution histogram revealing the spread of compensation across all employees, average salary comparisons by department showing IT leading at \$81,333, and employee count distribution indicating IT as the largest department with 15 employees.

Advanced Analytics The second screen presents correlation analysis between salary and years of service by department, hiring timeline trends from 2018-2021 showing departmental growth patterns, and a top 10 highest-paid employees ranking with Marcus Allen leading at \$99,000.

Detailed Data Tables The third screen provides tabular views of the processed data, including department performance metrics showing Finance with the highest average salary (\$64,714) and total costs, alongside IT having the most employees and highest total salary expenditure (\$1.22 million).

Export Functionality The dashboard includes data export options for Department Summary, Job Analysis, and Filtered Data, enabling users to download specific datasets for further analysis or reporting purposes.

This comprehensive dashboard effectively translates the backend data processing pipeline into actionable HR insights for strategic workforce management and compensation analysis.

□ Usage Instructions

Running the Data Pipeline

1. Single Step Execution:

```
python step1_load_inspect.py
```

```
python step2_clean_standardize.py
```

```
# continue with other steps
```

2. Complete Pipeline:

```
python step7_all_exports.py
```

Launching the Dashboard

1. Start the Dashboard:

```
streamlit run dashboard.py
```

2. Access the Application:

- Open browser to <http://localhost:8501>
- Dashboard loads automatically

3. Using the Dashboard:

- Select filters from sidebar
- Explore different visualizations
- Navigate between data tables using tabs
- Download data using export button

□ Troubleshooting Common Issues

The common issues you can face while running the files are listed here

Data Loading Errors:

Problem: FileNotFoundError: employees.csv not found

Solution: Ensure employees.csv is in the project directory

- Verify file name spelling and extension
- Check file permissions

Missing Dependencies

Problem: ModuleNotFoundError: No module named 'streamlit'

Solution: pip install streamlit pandas matplotlib numpy

Dashboard Import Errors

Problem: ImportError: DLL load failed

Solution: Use the simple dashboard version, Avoid plotly dependencies, Restart terminal

Empty Data Results

Problem: Dashboard shows no data

Solution: Run the complete data pipeline first, Verify CSV files are generated, Check data file paths

Bibliography

- Pandas Documentation: <https://pandas.pydata.org/docs/>
- NumPy Documentation: <https://numpy.org/doc/>
- Matplotlib Documentation:
<https://matplotlib.org/stable/contents.html>
- Plotly Python Graphing Library: <https://plotly.com/python/>
- Streamlit Documentation: <https://docs.streamlit.io/>
- Python Official Documentation: <https://docs.python.org/3/>