# **Employee Salary Analysis**

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Date: 11th March 2025

# **Employee Salary Analysis Report**

## Introduction

This report provides an analysis of employee salary data using Python. The project explores salary distributions, identifies outliers, and examines correlations between salary and other factors like department and experience.

## **Technologies Used**

- Python
- Pandas
- NumPy
- Matplotlib
- Seaborn

#### **Libraries Used**

#### **Pandas**

Pandas is used for data manipulation and analysis. It provides data structures like DataFrames to efficiently handle large datasets.

Example Code:

import pandas as pd
df = pd.read\_csv("employee\_data.csv")

## **NumPy**

NumPy is used for numerical computations, including handling arrays and mathematical functions.

Example Code:

import numpy as np

## **Matplotlib**

Matplotlib is used for creating static visualizations such as histograms and boxplots.

Example Code:

import matplotlib.pyplot as plt
plt.hist(df['Salary'], bins=30)
plt.show()

## Seaborn

Seaborn is used for advanced visualizations, including correlation heatmaps and KDE plots.

Example Code:

```
import seaborn as sns
sns.histplot(df['Salary'], bins=30, kde=True)
plt.show()
```

#### **Data Overview**

The dataset contains employee salary information, including:

- Department (Categorical)
- Experience (Numerical)
- Job Title (Categorical)
- Salary (Numerical)

## **Data Processing**

```
Handling Missing Values:
```

```
print(df.isnull().sum())
df = df.dropna(subset=['Salary'])
```

## **Data Visualization**

Salary Distribution:

```
plt.figure(figsize=(8, 5))
sns.histplot(df['Salary'], bins=30, kde=True, color='blue')
plt.title("Salary Distribution")
plt.xlabel("Salary")
plt.ylabel("Frequency")
plt.show()
```

Boxplot for Outlier Detection:

```
plt.figure(figsize=(8, 5))
sns.boxplot(x=df['Salary'], color='red')
plt.title("Salary Outliers")
plt.show()
```

# **Correlation Analysis**

The correlation between different numerical factors was examined using a heatmap:

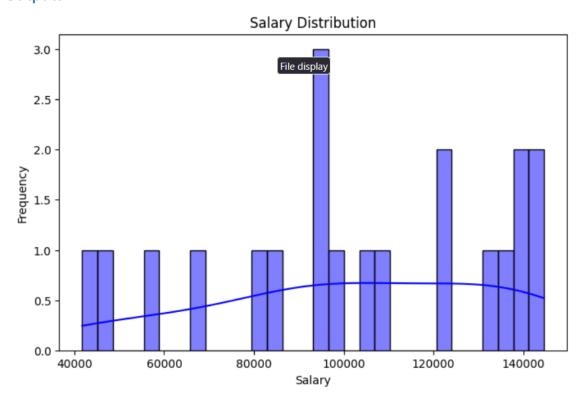
```
plt.figure(figsize=(10, 6))
sns.heatmap(df.corr(), annot=True, cmap="coolwarm", fmt=".2f", linewidths=0.5)
plt.title("Correlation Matrix")
plt.show()
```

## **Insights**

- The salary distribution shows a skewed pattern, indicating a higher concentration in certain salary ranges.

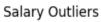
- Outliers in the salary data suggest significant variations, possibly due to high-level executive salaries.
- Correlation analysis helps understand relationships between salary and other factors like experience.

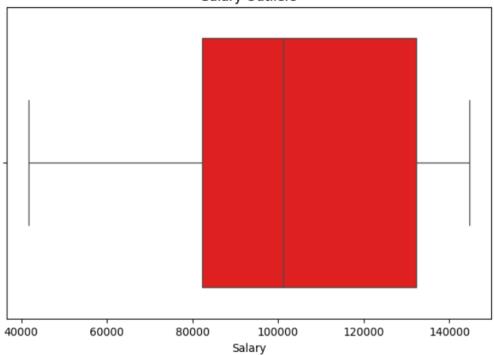
# **Outputs**





[1]: from google.colab import files uploaded = files.upload()





# **Conclusion**

This analysis provides valuable insights into employee salary trends, helping businesses understand pay structures and identify discrepancies.

# **References**

- 1. Pandas Documentation: https://pandas.pydata.org/docs/
- 2. NumPy Documentation: https://numpy.org/doc/
- 3. Matplotlib Documentation: https://matplotlib.org/stable/contents.html
- 4. Seaborn Documentation: https://seaborn.pydata.org/