3 Prac

1. Mean (Average)

- **Definition**: The sum of all values divided by the total number of values.
- Formula:

 $Mean = \sum xin \cdot \{Mean\} = \frac{x_i}{n} Mean = n\sum xi$

Where xix_ixi are the data points and nnn is the number of data points.

- Use: Gives a central value of the dataset.
- **Example**: If the salaries are ₹10000, ₹20000, and ₹30000,

$$\label{eq:mean} \begin{split} \mbox{Mean=10000+20000+300003=$20000} \\ \mbox{text{Mean}} &= \mbox{frac{10000+20000+30000}{3} = $20000 \\ \mbox{Mean=310000+20000+30000} \\ \mbox{$$?$} \end{aligned}$$

2. Median

- **Definition**: The middle value when the data is arranged in order.
- Use: It's a better measure than mean when data has outliers (extremely high or low values).
- **Example**: Salaries = ₹10000, ₹20000, ₹30000 → **Median** = ₹20000 Salaries = ₹10000, ₹20000, ₹1000000 → **Median** = ₹20000 (mean would be misleading here)

3. Minimum

- **Definition**: The **smallest** value in the dataset.
- **Use**: Shows the lower boundary of the data.
- Example: If ages are 21, 23, 25, min = 21

4. Maximum

- **Definition**: The **largest** value in the dataset.
- **Use**: Shows the upper boundary of the data.
- Example: If incomes are ₹25000, ₹50000, ₹90000 → max = ₹90000

5. Standard Deviation (std)

- **Definition**: A measure of **how spread out** the values are from the mean.
- Formula:

 $std=1n\sum_{i=1}^{n}(x_i-mean)^2 + \frac{1}{n}\sum_{i=1}^{n}(x_i-mean)^2 + \frac{1}{n}\sum_{i=1}^{n}(x_i-mean)^2$

• Use:

- Low std = data points are close to the mean
- High std = data points are spread out

• Example:

- o Salaries = ₹10000, ₹10500, ₹11000 → std is **low**
- o Salaries = ₹10000, ₹50000, ₹1000000 → std is **high**

Let me know if you'd like a visual plot to understand these better!

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You said:

what is data points

ChatGPT said:

Great question, Omkar!

Data Points – simply put:

A data point is a single value or observation in your dataset.

Example:

If you have student scores:

csharp

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[85, 90, 78, 92, 88]

Each number (85, 90, 78, etc.) is a data point. So this dataset has 5 data points.

In real-world terms:

- If you're collecting heights of 10 people → each person's height is a data point.
- If you're analyzing rainfall each day for a month → each day's rainfall is a data point.

***** Why it's important:

We use data points to:

- Calculate mean, median, standard deviation, etc.
- Visualize data (plots, graphs)
- Detect patterns, trends, or outliers

Would you like a simple chart or visual to show how data points form a dataset?

Objective

You're performing **descriptive statistics** (measures of central tendency and variability) on two datasets:

- 1. Employee_Salary_Dataset.csv Grouping numerical variables (like salary) by categorical (like age).
- 2. Iris.csv Grouping flower features (sepal, petal dimensions) by species and computing statistics.

Part 1: Employee Salary Dataset Analysis

Read and Display Dataset

python
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df = pd.read_csv('datasets/Employee_Salary_Dataset.csv')

df.head()

- Loads employee salary data into a DataFrame.
- Columns: ID, Experience_Years, Age, Gender, Salary.

Basic Descriptive Statistics

python

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print(df.mean(numeric_only=True))

print(df.median(numeric_only=True))

print(df.min(numeric_only=True))

print(df.max(numeric_only=True))

print(df.std(numeric_only=True))

print(df.describe(include='all'))

print(df.corr(numeric_only=True))

These lines compute:

- Mean, Median, Min, Max, Std. Deviation for numeric columns.
- describe() gives a full summary of stats including count, percentiles, and unique value counts.
- corr() shows correlation between numerical variables. For example:
 - o Experience_Years and Age have **strong positive correlation** (0.97), as expected.

Grouped Summary Statistics

python

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grouped = df.groupby('Age')['Salary'].agg(['mean', 'median', 'min', 'max', 'std'])
grouped_list = df.groupby('Age')['Salary'].apply(list)

- Groups Salary by Age, and computes:
 - o Mean salary per age, median, min, max, and standard deviation.
- Also builds a list of all salaries for each age (useful for plotting or further analysis).

This helps answer questions like: "What is the average salary of all 29-year-olds?"

Part 2: Iris Dataset Analysis

Load Iris Dataset

python

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dff = pd.read_csv('datasets/Iris.csv')

iris = dff.groupby('Species')

- Loads the classic **Iris dataset**, which contains:
 - o Features like SepalLengthCm, PetalWidthCm, etc.
 - o Categorical class: Species Iris-setosa, Iris-versicolor, Iris-virginica.

Descriptive Statistics by Species

python

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iris.describe(percentiles=[.25, .5, .75])

iris.mean(numeric_only=True)

iris.median(numeric_only=True)

iris.quantile(numeric_only=True)

iris.std(numeric_only=True)

iris.min(numeric_only=True)

iris.max(numeric_only=True)

This part answers:

- What are the average sepal/petal sizes per species?
- What are the minimum/maximum/median values?
- What's the **spread** (standard deviation)?
- How different are Iris-setosa vs. Iris-virginica in terms of petal length?

The describe() command gives a concise statistical summary with percentiles.

© Summary: What You've Achieved

Dataset	Action	Result
Employee Salary	Summarized age-wise salary stats	Insight into salary trends across ages
Employee Salary	Correlation between age/experience/salary	Age and experience are highly correlated with salary
Iris	Computed species-wise statistics	Understand how flower features vary across species

Would you like help plotting any of these statistics as graphs (e.g., boxplots or histograms)?

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You said:

explain code

ChatGPT said:

Sure Omkar! Let me walk you through the code you've written step by step. Your code is divided into **two major parts**:

Part 1: Descriptive Statistics on Employee Dataset



python

```
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```

import pandas as pd

import numpy as np

import matplotlib.pyplot as plt

import seaborn as sns

df = pd.read_csv('datasets/Employee_Salary_Dataset.csv')

df.head()

- pandas, numpy, matplotlib, and seaborn are imported for data handling and visualization.
- df.head() shows the first 5 rows of the dataset.

2. Basic Summary Statistics

python

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print("Mean")

print(df.mean(numeric_only=True))

- mean(), median(), min(), max(), std() are used to compute basic statistics only for numeric columns.
- describe() gives a full statistical summary.
- corr() calculates correlation between numeric columns.

☑ 3. Grouped Summary Statistics by Age

python

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grouped = df.groupby('Age')['Salary'].agg(['mean', 'median', 'min', 'max', 'std'])

Groups the dataset by Age and computes summary statistics for Salary.

python

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grouped_list = df.groupby('Age')['Salary'].apply(list)

• Converts each group into a list of Salary values per age.

Part 2: Descriptive Statistics on Iris Dataset

✓ 1. Load Iris Dataset

python

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dff = pd.read_csv('datasets/Iris.csv')

dff.head()

• Loads the classic Iris flower dataset.

2. Group by Species

python

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iris = dff.groupby('Species')

• Grouping the dataset by **Species**: Iris-setosa, Iris-versicolor, Iris-virginica.

3. Statistics by Species

Each of these provides different descriptive stats per species:

python

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print(iris.describe(percentiles=[.25, .5, .75]))

• Percentiles, mean, std, min, max for each numeric column.

python

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print(iris.mean(numeric_only=True))

print(iris.median(numeric_only=True))

print(iris.quantile(numeric_only=True))

• These give **mean**, **median**, and **quantiles** per species.

python

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print(iris.std(numeric_only=True))

print(iris.min(numeric_only=True))

print(iris.max(numeric_only=True))

• Shows standard deviation, min, and max per species.

python

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print(iris.describe(include='all'))

• Gives a complete summary for **each group** (species).

Summary

You successfully:

- 1. Computed overall and grouped statistics on the **Employee dataset**.
- 2. Computed detailed per-species descriptive stats from the **Iris dataset**.

This covers measures of central tendency (mean, median) and variability (std, min, max, percentiles) — exactly what the practical asked for.

Would you like me to help visualize any of these results using plots or graphs?

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