

Data Science

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DATA OPERATIONS

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Data Operations

Data Operations refers to the set of actions we perform to **understand, analyze, and transform** the data once it is collected and cleaned.

1. Reading

Reading data means **loading a dataset from an external source** into your Python environment using the pandas library.

We mostly use:

`pd.read_csv()` – for CSV files

`pd.read_excel()` – for Excel files

`pd.read_json()` – for JSON files

`pd.read_html()` – to extract tables from websites

General Codes:

Code 1: Reading a CSV file

```
import pandas as pd

df = pd.read_csv("students.csv")
print(df.head())
```

NOTE:

Loads the CSV file named students.csv into a DataFrame and displays the first 5 rows using .head().

Code 2: Reading an Excel file

```
df_excel = pd.read_excel("sales_data.xlsx")  
print(df_excel.info())
```

NOTE:

Reads an Excel file and shows structure: columns, datatypes, missing values.

Code 3: Reading a Dataset from URL

```
url = "https://raw.githubusercontent.com/mwaskom/seaborn-  
data/master/iris.csv"  
df_iris = pd.read_csv(url)  
print(df_iris.head())
```

NOTE:

Fetches a public CSV dataset directly from a URL.

Code 4: Reading a JSON file

```
df_json = pd.read_json("data.json")  
print(df_json.head())
```

NOTE:

Reads a JSON file. JSON is commonly used in APIs.

How to View Data

`.head(),`

`.tail()`

`.info()`

`.shape`

Task to do

Q. What is Custom delimiter ?

Q. Give the Code ?

2. Selecting

Code 1 : Selecting Columns **using df[]**

<code>df["Student"]</code>	<code># Single column</code>
<code>df[["Student", "CGPA"]]</code>	<code># Multiple columns</code>

NOTE:

`df['col']` gives a **Series**

`df[['col1', 'col2']]` gives a **DataFrame**

Code 2 : Selecting by Labels with .loc[]

Syntax: `df.loc[row_label, column_label]`

`df.loc[2]` # Entire row with index 2

`df.loc[0:2, "Student"]` # Student names from row 0 to 2

`df.loc[1:3, ["Student", "CGPA"]]`

Rows 1 to 3 with two columns (include 1,2,3)

Code 3 : Selecting by Index Numbers with .iloc[]

Syntax: `df.iloc[row_index, column_index]`

`df.iloc[2]` # Row at position 2

`df.iloc[1:4, [0, 2]]` # Rows 1 to 3, columns 0 and 2

or

`print(df.iloc[0:2, 0:2])` # Select first two rows and first two columns

`df.iloc[:, 1]` # All rows, column 1

Task	Code	Returns
Select one column	<code>df['Name']</code>	Series
Select multiple columns	<code>df[['Name', 'City']]</code>	DataFrame
Select row by label	<code>df.loc[1]</code>	Series
Select row and specific columns	<code>df.loc[0, ['Name', 'City']]</code>	Series
Select rows/cols by label	<code>df.loc[[0, 2], ['Name', 'Age']]</code>	DataFrame
Select row by position	<code>df.iloc[1]</code>	Series
Select value by position	<code>df.iloc[0, 1]</code>	Scalar value
Select range of rows and columns	<code>df.iloc[0:2, 0:2]</code>	DataFrame

3. Data Filtering

Example :

```
import pandas as pd
```

```
data = {'Name': ['Alice', 'Bob', 'Charlie', 'David'],  
        'Age': [25, 30, 35, 28],  
        'City': ['Delhi', 'Mumbai', 'Chennai', 'Delhi']}
```

```
df = pd.DataFrame(data)
```

```
print(df)
```

OUTPUT :

	Name	Age	City
0	Alice	25	Delhi
1	Bob	30	Mumbai
2	Charlie	35	Chennai
3	David	28	Delhi

1. Filter rows based on a single condition

CODE:

```
# Filter where Age > 30  
print(df[df['Age'] > 30])
```

OUTPUT:

	Name	Age	City
2	Charlie	35	Chennai

2. Filter rows based on multiple conditions

Use & (and), | (or), ~ (not) with parentheses ()

CODE:

```
# Age > 25 AND City is Delhi  
print(df[(df['Age'] > 25) & (df['City'] == 'Delhi')])
```

OUTPUT:

	Name	Age	City
3	David	28	Delhi

Why use .isin()?

- In Pandas, **.isin()** is used to filter rows where a column's value is **in a list or set of multiple values**.

Eg: `df[df['City'].isin(['Delhi', 'Mumbai', 'Chennai'])]`

Without .isin():

- We need to write **multiple OR conditions**, which gets messy:

Eg: `df[(df['City'] == 'Delhi') | (df['City'] == 'Mumbai') | (df['City'] == 'Chennai')]`

Key Benefits of .isin():

Advantage	Explanation
◆ Easy syntax	More readable than writing multiple == and `
◆ Works with lists/sets	You can filter using many values at once
◆ Works with NOT using ~	Like ~df['City'].isin(['Delhi']) to exclude Delhi

3. Filter using string methods

CODE:

```
# Filter where Name starts with 'A'
```

```
print(df[df['Name'].str.startswith('A')])
```

OUTPUT:

	Name	Age	City
0	Alice	25	Delhi

Quiz

```
import pandas as pd
```

```
data = {'Name': ['Alice', 'Bob', 'Charlie', 'David', 'Eva'],  
        'Age': [25, 30, 35, 28, 22],  
        'City': ['Delhi', 'Mumbai', 'Chennai', 'Delhi',  
                 'Kolkata']}  
df = pd.DataFrame(data)
```

Q1. Filter all rows where the Age is greater than 28

Write the filtering line of code.

Q2. Filter all rows where the City is either Delhi or Kolkata

Use `.isin()` method.

Q3. Filter all rows where Name starts with the letter 'C'

Q4. Filter all rows where Age is NOT 30 or 35

Q5. Filter all rows where City is not Delhi, using the ~ operator.

4. Data Manipulation

- **Manipulating data** means **modifying the contents** of a DataFrame — like changing values, adding/removing columns, renaming, replacing, or applying calculations.

This includes:

- Adding new columns
- Updating existing values
- Replacing missing or incorrect data
- Changing data types
- Applying operations to transform data

Common Manipulation Operations:

S.No.	Operation	Example
1	Add new column	Add Total or Grade column
2	Modify values	Update names, marks, etc.
3	Apply formula	Convert marks to percentage
4	Replace values	Replace 'NaN' or wrong data
5	Rename columns	Make headers cleaner or standardized

Example 1: Create a Sample DataFrame

CODE:

```
import pandas as pd
```

```
data = {'Name': ['Alice', 'Bob', 'Charlie'],  
        'Marks': [85, 90, 78],  
        'City': ['Delhi', 'Mumbai', 'Chennai']}
```

```
df = pd.DataFrame(data)
```

```
print(df)
```

OUTPUT:

	Name	Marks	City
0	Alice	85	Delhi
1	Bob	90	Mumbai
2	Charlie	78	Chennai

Code1: Add a new column (Let's add a column Grade based on Marks)

CODE:

```
df['Grade'] = ['B', 'A', 'C']  
print(df)
```

OUTPUT:

	Name	Marks	City	Grade
0	Alice	85	Delhi	B
1	Bob	90	Mumbai	A
2	Charlie	78	Chennai	C

Code 2: Update values in a column (Let's say Bob improved his marks to 95)

CODE:

```
df.loc[df['Name'] == 'Bob', 'Marks'] = 95  
print(df)
```

OUTPUT:

	Name	Marks	City	Grade
0	Alice	85	Delhi	B
1	Bob	95	Mumbai	A
2	Charlie	78	Chennai	C

Code 3: Apply a function to transform data (Let's add 5 bonus marks to each student)

CODE:

```
df['Final_Marks'] = df['Marks'].apply(lambda x: x + 5)
```

OUTPUT:

	Name	Marks	City	Grade	Final_Marks
0	Alice	85	Delhi	B	90
1	Bob	95	Mumbai	A	100
2	Charlie	78	Chennai	C	83

Or

CODE:

```
def add_bonus(x):
```

```
    return x + 5
```

```
df['Final_Marks'] = df['Marks'].apply(add_bonus)
```

NOTE:

- (1) Lambda is a Short, inline function.
- (2) No need to define a separate named function.

Code 4: Replace values (Let's change City = 'Chennai' to 'Madras')

CODE:

```
df['City'] = df['City'].replace('Chennai', 'Madras')  
print(df)
```

OUTPUT:

	Name	Marks	City	Grade	Final_Marks
0	Alice	85	Delhi	B	90
1	Bob	95	Mumbai	A	100
2	Charlie	78	Madras	C	83

Code 5: Change Data Type (Change Marks from int to float)

CODE:

```
df['Marks'] = df['Marks'].astype(float)
print(df.dtypes)
```

OUTPUT:

Name	object
Marks	float64
City	object
Grade	object
Final_Marks	int64
dtype:	object

Summary

Task	Code
Add new column	<code>df['NewCol'] = [...]</code>
Update cell value	<code>df.loc[row, 'Col'] = new_value</code>
Apply function to column	<code>df['Col'].apply(func)</code>
Replace values	<code>df['Col'].replace(old, new)</code>
Change data type	<code>df['Col'].astype(new_type)</code>

Special Cases

Ex1: Change a specific value in between (e.g., add 5 to only the second row)

- Modify a specific cell (or row/column) using `.loc[]` or `.iloc[]`.

CODE: Add 5 to the Marks of the second student (index 1):

```
df.loc[1, 'Marks'] = df.loc[1, 'Marks'] + 5
```

OR

```
df.iloc[1, 1] = df.iloc[1, 1] + 5
```

(It's the 2nd row and 2nd column — index 1 and 1):

Ex 2: Check marks in float

```
print(df)
```

OUTPUT:

	Name	Marks	City	Grade
0	Alice	85.0	Delhi	B
1	Bob	100.0	Mumbai	A
2	Charlie	88.0	Madras	C

5. Data Sorting :

Sorting in Pandas means arranging your DataFrame based on the values of one or more columns.

It includes:

- Sort by column values (like Marks, Name, etc.)
- Sort by index
- Sort in ascending or descending order
- Sort multiple columns (e.g., by Subject, then Marks)

CODE:

```
df.sort_values(by='column_name', ascending=True/False)
```

Ex: Create a Data Set

```
import pandas as pd
```

```
data = {'Name': ['Alice', 'Bob', 'Charlie', 'David', 'Eva'],  
        'Marks': [85, 95, 78, 88, 95],  
        'Subject': ['Math', 'Physics', 'Chemistry', 'Math',  
                    'Physics']}  
df = pd.DataFrame(data)  
print(df)
```

	Name	Marks	Subject
0	Alice	85	Math
1	Bob	95	Physics
2	Charlie	78	Chemistry
3	David	88	Math
4	Eva	95	Physics

Code1. Sort by a single column (ascending)

Sort students by Marks in ascending order:

CODE:

```
df_sorted = df.sort_values(by='Marks')  
print(df_sorted)
```

OUTPUT:

	Name	Marks	Subject
2	Charlie	78	Chemistry
0	Alice	85	Math
3	David	88	Math
1	Bob	95	Physics
4	Eva	95	Physics

Code 2. Sort by a single column (descending)

```
df.sort_values(by='Marks', ascending=False)
```

Code 2. Sort by multiple columns

Sort first by Subject, then by Marks in descending order:

CODE:

```
df.sort_values(by=['Subject', 'Marks'], ascending=[True, False])
```

OUTPUT:

	Name	Marks	Subject
2	Charlie	78	Chemistry
0	Alice	85	Math
3	David	88	Math
1	Bob	95	Physics
4	Eva	95	Physics

Code 3. Sort by index

CODE:

```
import pandas as pd

data = {
    'Name': ['Alice', 'Bob', 'Charlie'],
    'Marks': [85, 95, 78]
}

df = pd.DataFrame(data, index=[2, 0, 1])
print("Original DataFrame:")
print(df)
```

OUTPUT:

	Name	Marks
2	Alice	85
0	Bob	95
1	Charlie	78

Now shuffle the data

CODE:

```
df_sorted = df.sort_index()  
print("Sorted by index:")  
print(df_sorted)
```

OUTPUT:

	Name	Marks
0	Bob	95
1	Charlie	78
2	Alice	85

General Syntax

`df.sort_index(ascending=True) # Ascending (default)`

`df.sort_index(ascending=False) # Descending`

6. Data Grouping :

Grouping means **splitting** your data into groups based on some category and then applying a **function** (like sum, mean, count) to each group.

Syntax of groupby ():

```
df.groupby('column_name')
```

With function:

```
df.groupby('column_name').sum()
```

```
df.groupby('column_name').mean()
```

```
df.groupby('column_name').count()
```


Example :

Name	Subject	Marks
Alice	Math	85
Bob	Math	90
Alice	Science	88
Bob	Science	84
Carol	Math	92

CODE:

```
import pandas as pd
```

```
data = {  
    'Name': ['Alice', 'Bob', 'Alice', 'Bob', 'Carol'],  
    'Subject': ['Math', 'Math', 'Science', 'Science', 'Math'],  
    'Marks': [85, 90, 88, 84, 92]  
}
```

```
df = pd.DataFrame(data)  
print(df)
```

OUTPUT:

	Name	Subject	Marks
0	Alice	Math	85
1	Bob	Math	90
2	Alice	Science	88
3	Bob	Science	84
4	Carol	Math	92

Code 1. Total Marks by Student

```
df.groupby('Name')['Marks'].sum()
```

OUTPUT:

Name

Alice 173

Bob 174

Carol 92

Name: Marks, dtype: int64

Code 2. Average Marks per Subject

```
df.groupby('Subject')['Marks'].mean()
```

OUTPUT:

Subject

Math 89.0

Science 86.0

Name: Marks, dtype: float64

Multiple Grouping Columns

CODE 1 :

```
df.groupby(['Name', 'Subject'])['Marks'].mean()
```

OUTPUT:

Name	Subject	
Alice	Math	85
	Science	88
Bob	Math	90
	Science	84
Carol	Math	92

Name: Marks, dtype: int64

Code 2 : For Duplicate values

```
data = {  
    'Name': ['Alice', 'Alice', 'Bob', 'Bob', 'Carol'],  
    'Subject': ['Math', 'Math', 'Math', 'Science', 'Math'],  
    'Marks': [85, 95, 90, 84, 92]  
}  
df = pd.DataFrame(data)  
df.groupby(['Name', 'Subject'])['Marks'].mean()  
Or
```

OUTPUT:

Name	Subject	
Alice	Math	90.0 # (85+95)/2
Bob	Math	90.0
	Science	84.0
Carol	Math	92.0

Name: Marks, dtype: float64

Code 3. Use Unstack : Convert row index → columns

```
grouped = df.groupby(['Name', 'Subject'])['Marks'].mean()
print(grouped)
print(grouped.unstack())
```

OUTPUT:

Subject	Math	Science
Alice	90.0	NaN
Bob	90.0	84.0
Carol	92.0	NaN

Code 4. Use of reset index

Again back to original data

```
print(grouped.reset_index())
```

7. Rearranging

Rearranging in Pandas — this includes changing the layout, column orders, or reshaping your DataFrame.

It means **changing the order** of:

- Columns
- Rows
- Or even reshaping the structure of your DataFrame.

Code1. Rearranging Columns

(Rearrange columns by simply changing their order)

```
import pandas as pd
df = pd.DataFrame({
    'Name': ['Alice', 'Bob', 'Charlie'],
    'Age': [25, 30, 22],
    'City': ['Delhi', 'Mumbai', 'Chennai']
})

# Reorder columns: move 'City' first
df_rearranged = df[['City', 'Name', 'Age']]
print(df_rearranged)
```

OUTPUT:

	City	Name	Age
0	Delhi	Alice	25
1	Mumbai	Bob	30
2	Chennai	Charlie	22

Code2. Rearranging Rows by Index

Reverse the order of rows using slicing like `df[::-1]`.

```
import pandas as pd
```

```
df = pd.DataFrame({  
    'Name': ['Alice', 'Bob', 'Charlie'],  
    'Age': [25, 30, 22],  
    'City': ['Delhi', 'Mumbai', 'Chennai']  
})
```

Reverse row order

```
df_reversed = df[::-1]  
print(df_reversed)
```

OUTPUT:

	Name	Age	City
2	Charlie	22	Chennai
1	Bob	30	Mumbai
0	Alice	25	Delhi

Code 3. Rearranging Rows Using .reindex()

reorder rows manually

```
# Rearranging rows using custom index
```

```
df_custom = df.reindex([2, 0, 1])
```

```
print(df_custom)
```

OUTPUT:

	Name	Age	City
2	Charlie	22	Chennai
0	Alice	25	Delhi
1	Bob	30	Mumbai

Code 4. Transposing the DataFrame with .T

```
# Transpose the DataFrame
```

```
print(df.T)
```

OUTPUT:

	0	1	2
Name	Alice	Bob	Charlie
Age	25	30	22
City	Delhi	Mumbai	Chennai

8. Ranking

- In Pandas, `.rank()` is used to assign ranks to data values based on their size — smallest gets rank 1 by default.
- It's often used in grading, competition scores, sales analysis, etc.

Syntax:

```
DataFrame['column'].rank(method='average',ascending=True)
```

Parameters

Parameter	Description
method	How to assign ranks to equal values: 'average', 'min', 'max', 'dense', 'first'
ascending	True for smallest rank = 1, False for highest = 1
axis	Use 0 for columns, 1 for rows (default is column-wise ranking)

Code 1. Ranking Exam Scores

```
import pandas as pd
```

```
df = pd.DataFrame({  
    'Name': ['A', 'B', 'C', 'D', 'E'],  
    'Score': [85, 92, 88, 92, 70]  
})
```

```
df['Rank'] = df['Score'].rank(ascending=False)  
print(df)
```

OUTPUT:

	Name	Score	Rank
0	A	85	4.0
1	B	92	1.5
2	C	88	3.0
3	D	92	1.5
4	E	70	5.0

Code 2. Average, min, max, dense

```
import pandas as pd

df = pd.DataFrame({
    'Name': ['A', 'B', 'C', 'D', 'E'],
    'Score': [90, 80, 90, 70, 60]
})
```

```
df['Rank_average'] = df['Score'].rank(method='average',  
ascending=False)
```

```
df['Rank_min'] = df['Score'].rank(method='min',  
ascending=False)
```

```
df['Rank_max'] = df['Score'].rank(method='max',  
ascending=False)
```

```
df['Rank_dense'] = df['Score'].rank(method='dense',  
ascending=False)
```

```
print(df)
```

OUTPUT:

	Name	Score	Rank_average	Rank_min	Rank_max	Rank_dense
0	A	90	1.5	1.0	2.0	1.0
1	B	80	3.0	3.0	3.0	2.0
2	C	90	1.5	1.0	2.0	1.0
3	D	70	4.0	4.0	4.0	3.0
4	E	60	5.0	5.0	5.0	4.0

Code 3. Axis=1 (Row-wise Ranking):

```
df2 = pd.DataFrame({  
    'Math': [90, 40, 70],  
    'Science': [80, 60, 75],  
    'English': [85, 55, 65]  
}, index=['Student1', 'Student2', 'Student3'])
```

Row-wise:

```
rank_rowwise = df2.rank(axis=1, ascending=False)  
print(rank_rowwise)
```

Column-wise:

```
rank_colwise = df2.rank(axis=0, ascending=False)  
print(rank_colwise)
```

