

Course Name:	Data Analysis Laboratory (216H03L501)	Semester:	V
Date of Performance:	14 / 07 / 2025	DIV/ Batch No:	HDA2
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Experiment No: 1

Title: Studies on Pandas library of python.

Objectives of the Experiment:

1. To understand and apply the fundamental functionalities of the pandas library for data analysis.
2. To manipulate and transform datasets using filtering, sorting, and column operations.
3. To analyze data using grouping and aggregation techniques to derive meaningful insights.

COs to be achieved:

CO1: Understand basic concepts of data analytics to solve real-world problems

Books/ Journals/ Websites referred:

1. <https://www.kaggle.com/datasets>
2. <https://www.w3schools.com/python/pandas>

Theory:

Pandas is a Python library used for working with data sets.
It has functions for analyzing, cleaning, exploring, and manipulating data.

Problem statement/ Tasks

Task 1: Import Required Libraries and Dataset

- Import pandas and load a real-world CSV dataset (e.g., Titanic, Student Performance, COVID-19).
- Display the first and last 5 records using head() and tail().

Task 2: Basic Exploration of the Dataset

- Display the dataset shape using .shape, column names using .columns, and data types using .dtypes.
- Generate summary statistics using .describe() and data info using .info().

Task 3: Identify Missing and Duplicate Data

- Detect missing values using .isnull().sum().
- Remove or fill missing values using .dropna() or .fillna().

- Check for and remove duplicate rows using `.duplicated()` and `.drop_duplicates()`.

Task 4: Filtering Records

- Extract rows based on specific conditions (e.g., students who scored more than 80%, passengers who survived).

Task 5: Sorting the Dataset

- Sort the dataset based on one or more columns using `.sort_values()`.
 - Example: Sort by age or total score.

Task 6: Creating or Modifying Columns

- Create new columns from existing ones (e.g., Total Marks = Math + Science + English).
- Drop unnecessary columns using `.drop()`.
- Rename columns using `.rename()`.

Task 7: Grouping and Aggregation

- Use `.groupby()` to find average, count, or sum based on a categorical column.
- Example:
 - Average marks by gender: `df.groupby('Gender')['Marks'].mean()`
 - Survival rate by class: `df.groupby('Pclass')['Survived'].mean()`

Task 8: Pivot Tables or Multi-Level Grouping (Optional for advanced students)

- Create pivot tables using `.pivot_table()` to summarize complex data.
 - Example: Average score by gender and class.

Task 9: Insight Generation

- Write 3-5 key insights based on the group-by and aggregated data.
- Example:
 - "Female students have higher average marks in English."
 - "Survival rate is highest for first-class passengers."

Code:

```
[ ] # Print the version
    print(f"pandas version: {pd.__version__}")

⇒ pandas version: 2.2.2
```

```
csp = pd.read_csv(os.path.join(path, "college_student_placement_dataset.csv"))
csp
```

	College_ID	IQ	Prev_Sem_Result	CGPA	Academic_Performance	Internship_Experience	Extra_Curricular_Score	Communication_Skills	Projects_Completed	Placement
0	CLG0030	107	6.61	6.28	8	No	8	8	4	No
1	CLG0061	97	5.52	5.37	8	No	7	8	0	No
2	CLG0036	109	5.36	5.83	9	No	3	1	1	No
3	CLG0055	122	5.47	5.75	6	Yes	1	6	1	No
4	CLG0004	96	7.91	7.69	7	No	8	10	2	No
...
9995	CLG0021	119	8.41	8.29	4	No	1	8	0	Yes
9996	CLG0098	70	9.25	9.34	7	No	0	7	2	No
9997	CLG0066	89	6.08	6.25	3	Yes	3	9	5	No
9998	CLG0045	107	8.77	8.92	3	No	7	5	1	No
9999	CLG0060	109	9.41	9.77	8	No	3	5	5	No

10000 rows x 10 columns

```
csp.describe()
```

	IQ	Prev_Sem_Result	CGPA	Academic_Performance	Extra_Curricular_Score	Communication_Skills	Projects_Completed
count	10000.000000	10000.000000	10000.000000	10000.000000	10000.000000	10000.000000	10000.000000
mean	99.471800	7.535673	7.532379	5.546400	4.970900	5.561800	2.513400
std	15.053101	1.447519	1.470141	2.873477	3.160103	2.900866	1.715959
min	41.000000	5.000000	4.540000	1.000000	0.000000	1.000000	0.000000
25%	89.000000	6.290000	6.290000	3.000000	2.000000	3.000000	1.000000
50%	99.000000	7.560000	7.550000	6.000000	5.000000	6.000000	3.000000
75%	110.000000	8.790000	8.770000	8.000000	8.000000	8.000000	4.000000
max	158.000000	10.000000	10.460000	10.000000	10.000000	10.000000	5.000000

```
csp.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10000 entries, 0 to 9999
Data columns (total 10 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   College_ID                            10000 non-null  object
1   IQ                                     10000 non-null  int64
2   Prev_Sem_Result                       10000 non-null  float64
3   CGPA                                  10000 non-null  float64
4   Academic_Performance                  10000 non-null  int64
5   Internship_Experience                 10000 non-null  object
6   Extra_Curricular_Score                10000 non-null  int64
7   Communication_Skills                  10000 non-null  int64
8   Projects_Completed                    10000 non-null  int64
9   Placement                             10000 non-null  object
dtypes: float64(2), int64(5), object(3)
memory usage: 781.4+ KB
```

```
[ ] print("Missing values:\n", csp.isnull().sum())
```

```
Missing values:  
College_ID      0  
IQ              0  
Prev_Sem_Result 0  
CGPA            0  
Academic_Performance 0  
Internship_Experience 0  
Extra_Curricular_Score 0  
Communication_Skills 0  
Projects_Completed 0  
Placement       0  
dtype: int64
```

```
csp['CGPA'].value_counts()
```

```
count  
CGPA  
9.41    35  
7.29    32  
6.72    32  
6.09    31  
9.47    31  
...     ...  
10.44    1  
4.57     1  
4.67     1  
4.56     1  
4.54     1  
590 rows × 1 columns  
dtype: int64
```

```
[ ] print("Total student=",csp['CGPA'].value_counts())
```

```
Total student= CGPA
9.41      35
7.29      32
6.72      32
6.09      31
9.47      31
..
10.44     1
4.57      1
4.67      1
4.56      1
4.54      1
Name: count, Length: 590, dtype: int64
```

```
[ ] csp['CGPA'].value_counts()
```

```
count
CGPA
9.41    35
7.29    32
6.72    32
6.09    31
9.47    31
...
10.44    1
4.57     1
4.67     1
4.56     1
4.54     1
590 rows x 1 columns

dtype: int64
```

```
csp['College_ID'].str.lower()
```



College_ID

0	clg0030
1	clg0061
2	clg0036
3	clg0055
4	clg0004
...	...
9995	clg0021
9996	clg0098
9997	clg0066
9998	clg0045
9999	clg0060

10000 rows × 1 columns

dtype: object

```
[ ] csp['Placement'].value_counts()
```



count

Placement

No	8341
Yes	1659

dtype: int64

```
[ ] print("Total placed = ",csp['Placement'].value_counts()['Yes'])
```



Total placed = 1659

```
print("First 5 records:")
print(csp.head())

print("\nLast 5 records:")
print(csp.tail())
```

First 5 records:

	College_ID	IQ	Prev_Sem_Result	CGPA	Academic_Performance	\
0	CLG0030	107	6.61	6.28	8	
1	CLG0061	97	5.52	5.37	8	
2	CLG0036	109	5.36	5.83	9	
3	CLG0055	122	5.47	5.75	6	
4	CLG0004	96	7.91	7.69	7	

	Internship_Experience	Extra_Curricular_Score	Communication_Skills	\
0	No	8	8	
1	No	7	8	
2	No	3	1	
3	Yes	1	6	
4	No	8	10	

	Projects_Completed	Placement
0	4	No
1	0	No
2	1	No
3	1	No
4	2	No

Last 5 records:

	College_ID	IQ	Prev_Sem_Result	CGPA	Academic_Performance	\
9995	CLG0021	119	8.41	8.29	4	
9996	CLG0098	70	9.25	9.34	7	
9997	CLG0066	89	6.08	6.25	3	
9998	CLG0045	107	8.77	8.92	3	
9999	CLG0060	109	9.41	9.77	8	

	Internship_Experience	Extra_Curricular_Score	Communication_Skills	\
9995	No	1	8	
9996	No	0	7	
9997	Yes	3	9	
9998	No	7	5	
9999	No	3	5	

	Projects_Completed	Placement
9995	0	Yes
9996	2	No
9997	5	No
9998	1	No
9999	5	No

```
print("Dataset Shape:", csp.shape)
print("\nColumn Names:", csp.columns)
print("\nData Types:\n", csp.dtypes)
```

Dataset Shape: (10000, 10)

Column Names: Index(['College_ID', 'IQ', 'Prev_Sem_Result', 'CGPA', 'Academic_Performance',
'Internship_Experience', 'Extra_Curricular_Score',
'Communication_Skills', 'Projects_Completed', 'Placement'],
dtype='object')

Data Types:

College_ID	object
IQ	int64
Prev_Sem_Result	float64
CGPA	float64
Academic_Performance	int64
Internship_Experience	object
Extra_Curricular_Score	int64
Communication_Skills	int64
Projects_Completed	int64
Placement	object
dtype:	object

```
csp.dropna(inplace=True)
print("\nMissing values after handling:\n", csp.isnull().sum())

print("\nNumber of duplicate rows:", csp.duplicated().sum())

csp.drop_duplicates(inplace=True)
print("\nDataset shape after removing duplicates:", csp.shape)
```

Missing values after handling:

College_ID	0
IQ	0
Prev_Sem_Result	0
CGPA	0
Academic_Performance	0
Internship_Experience	0
Extra_Curricular_Score	0
Communication_Skills	0
Projects_Completed	0
Placement	0
dtype:	int64

Number of duplicate rows: 0

Dataset shape after removing duplicates: (10000, 10)


```

high = csp[csp['CGPA'] > 8.0]
print("\nStudents who scored more than 8 CGPA:\n", high)

```

Students who scored more than 8 CGPA:

	College_ID	IQ	Prev_Sem_Result	CGPA	Academic_Performance	\
7	CLG0096	111	8.77	8.76	7	
9	CLG0057	108	8.82	8.60	4	
10	CLG0063	93	8.73	8.90	2	
12	CLG0064	103	8.64	9.01	7	
13	CLG0017	71	8.74	8.40	6	
...	
9994	CLG0064	117	8.71	8.44	6	
9995	CLG0021	119	8.41	8.29	4	
9996	CLG0098	70	9.25	9.34	7	
9998	CLG0045	107	8.77	8.92	3	
9999	CLG0060	109	9.41	9.77	8	

	Internship_Experience	Extra_Curricular_Score	Communication_Skills	\
7	No	3	1	
9	No	5	9	
10	Yes	5	6	
12	Yes	8	6	
13	No	0	5	
...	
9994	No	9	4	
9995	No	1	8	
9996	No	0	7	
9998	No	7	5	
9999	No	3	5	

	Projects_Completed	Placement
7	2	Yes
9	1	No
10	0	No
12	1	No
13	2	No
...
9994	4	Yes
9995	0	Yes
9996	2	No
9998	1	No
9999	5	No

[4088 rows x 10 columns]



```
csp_sorted = csp.sort_values(by='CGPA')
print("\nDataset sorted by CGPA:")
print(csp_sorted.head())
```



Dataset sorted by CGPA:

	College_ID	IQ	Prev_Sem_Result	CGPA	Academic_Performance	\
9333	CLG0044	113	5.03	4.54	6	
8835	CLG0017	99	5.01	4.56	5	
8574	CLG0025	97	5.06	4.57	6	
3332	CLG0005	123	5.02	4.58	3	
319	CLG0029	98	5.03	4.59	10	

	Internship_Experience	Extra_Curricular_Score	Communication_Skills	\
9333	Yes	9	5	
8835	Yes	0	9	
8574	No	7	5	
3332	Yes	2	9	
319	No	10	4	

	Projects_Completed	Placement
9333	4	No
8835	4	No
8574	2	No
3332	5	Yes
319	0	No



```
csp['Knowledge'] = csp['IQ'] + csp['CGPA'] + csp['Academic_Performance']

csp.rename(columns={'College_ID': 'University_ID', 'CGPA': 'Cumulative_Grade_Point_Average'}, inplace=True)

print("\nDataset with new column and renamed columns:")
print(csp.head())
```



Dataset with new column and renamed columns:

	University_ID	IQ	Prev_Sem_Result	Cumulative_Grade_Point_Average	\
0	CLG0030	107	6.61	6.28	
1	CLG0061	97	5.52	5.37	
2	CLG0036	109	5.36	5.83	
3	CLG0055	122	5.47	5.75	
4	CLG0004	96	7.91	7.69	

	Academic_Performance	Internship_Experience	Extra_Curricular_Score	\
0	8	No	8	
1	8	No	7	
2	9	No	3	
3	6	Yes	1	
4	7	No	8	

	Communication_Skills	Projects_Completed	Placement	Knowledge
0	8	4	No	121.28
1	8	0	No	110.37
2	1	1	No	123.83
3	6	1	No	133.75
4	10	2	No	110.69

```
print(csp.groupby('Internship_Experience')['Cumulative_Grade_Point_Average'].mean())

print("\nCount of students by Internship Experience:")
print(csp.groupby('Internship_Experience').size())

print("\nSum of Projects Completed by Internship Experience:")
print(csp.groupby('Internship_Experience')['Projects_Completed'].sum())
```

↔ Internship_Experience

No	7.547344
Yes	7.509591

Name: Cumulative_Grade_Point_Average, dtype: float64

Count of students by Internship Experience:

Internship_Experience	
No	6036
Yes	3964

dtype: int64

Sum of Projects Completed by Internship Experience:

Internship_Experience	
No	15257
Yes	9877

Name: Projects_Completed, dtype: int64

```
pivot_table_result = csp.pivot_table(
    values='Academic_Performance',
    index='IQ',
)
pivot_table_result
```

↔ Academic_Performance

IQ	
41	2.0
42	2.0
44	8.0
45	2.0
51	4.0
...	...
148	1.0
150	3.0
152	5.0
157	4.0
158	8.0

104 rows × 1 columns

Post Lab Subjective/Objective type Questions:

1. What is the difference between .info() and .describe() in pandas?

The .info() and .describe() functions in pandas serve different purposes when exploring a dataset. The .info() method provides a concise summary of the DataFrame, including the number of non-null entries, column names, data types, and memory usage. It is particularly useful for getting a quick overview of the dataset's structure and identifying missing values. On the other hand, the .describe() method generates descriptive statistics for numerical columns by default, such as count, mean, standard deviation, minimum and maximum. It helps in understanding the distribution and spread of the data. While .info() focuses on the structure and completeness of the data, .describe() is more concerned with the statistical properties and variability of the dataset.

2. How does pandas handle missing data? Mention at least two functions used for this purpose.

Pandas handles missing data using NaN (Not a Number) values. `df.isnull()` / `df.notnull()`: Identify missing data `df.dropna()`: Removes rows or columns with missing values.

3. What is a pivot table in pandas, and how is it useful in summarizing data?

A pivot table in pandas is a powerful tool that allows you to summarize and aggregate data using multiple dimensions which is created using `pd.pivot_table()`. It is Useful for quickly reshaping data, comparing subgroups, and generating report. Ex: Group sales data by Region and Product and calculate average sales.

4. What were the key insights you discovered from the dataset during your analysis?

During the analysis of the College Student Placement Factors dataset, several key insights emerged. Firstly, students with higher CGPAs and consistent academic performance across 10th and 12th grades were more likely to get placed. Secondly, participation in extracurricular activities and internships showed a positive correlation with placement outcomes, indicating the importance of soft skills and practical exposure. Additionally, students with strong communication and technical skills, as reflected in the dataset, had a higher placement rate. It was also observed that students from specific degree programs and colleges had better placement statistics, highlighting institutional influence. These insights can help institutions and students focus on the most impactful factors to improve employability.

Conclusion:

I have successfully used pandas, which involved exploring and analyzing data in Python. Through this experiment, I learned how to load real-world datasets, perform basic data exploration using functions like .info() and .describe(), handle missing and duplicate data, apply filtering, sorting. This experiment helped me understand the practical importance of data analysis in identifying trends and making informed decisions.

