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## **EXPERIMENT NO: 1**

**Aim**: Introduction to Data science and Data preparation using Pandas steps.

# Theory:

Data preparation is a fundamental step in data science, involving the cleaning and transformation of raw data into a structured and analyzable format. Pandas, a powerful Python library, provides efficient tools for handling missing values, encoding categorical data, and scaling numerical features. Proper preprocessing enhances dataset quality, ensuring consistency and reliability for further analysis and machine learning models.

### **Problem Statement:**

The Placement Data dataset contains various attributes related to students' academic performance, placement status, and salary packages. The objective of this experiment is to:

- Identify key trends in student placements based on academic performance.
- Analyze the distribution of salary packages.
- Handle missing data and remove inconsistencies.
- Standardize and normalize the data for further analysis.

By cleaning the placement dataset and applying data preprocessing steps, the goal is to improve data reliability, analyze student performance trends, and provide valuable insights for academic and recruitment decisions.

### **Dataset Overview:**

The dataset provides detailed information about student placements, academic performance, and salary distributions. It contains multiple columns, each capturing specific attributes related to students' educational backgrounds and employment outcomes. Below is a breakdown of the columns and their relevance: The dataset provides insights into student placements, containing columns such as:

- 1. StudentID: Unique identifier for each student.
- **2. CGPA:** Cumulative Grade Point Average of the student.
- **3. Internships:** Number of internships completed.
- **4. Projects:** Number of academic or industry projects undertaken.
- **5.** Workshops/Certifications: Number of workshops attended or certifications earned.
- **6. Aptitude Test Score:** Score obtained in the aptitude test.
- 7. Soft Skills Rating: Rating of soft skills on a predefined scale.
- **8. Extracurricular Activities:** Participation in extracurricular activities.
- **9. Placement Training:** Whether the student underwent placement training (Yes/No).
- **10. SSC Marks:** Secondary school exam scores.
- 11. HSC Marks: Higher secondary exam scores.

**12. Placement Status:** Whether the student was placed or not.

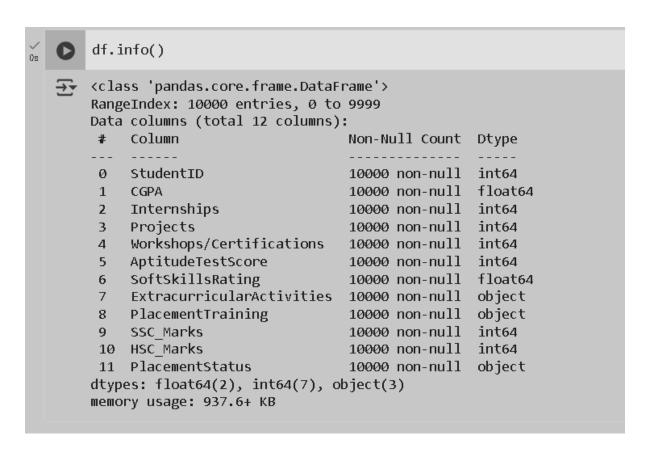
# **Steps:**

## **Loading The Dataset**



# **Description of the dataset**

### a. Information about dataset



### **b.** Description of Dataset

```
# Display basic information about the dataset
print("Dataset Information:")
print(df.info())

# Display summary statistics
print("\nDataset Description:")
print(df.describe())

# Display the first few rows of the dataset
print("\nFirst 5 Rows of the Dataset:")
print(df.head())
```

```
→ Dataset Information:
    <class 'pandas.core.frame.DataFrame'>
RangeIndex: 10000 entries, 0 to 9999
    Data columns (total 12 columns):
                                     Non-Null Count Dtype
     # Column
                                     10000 non-null int64
     0 StudentID
                                     10000 non-null
                                                     float64
         CGPA
         Internships
                                     10000 non-null
                                                     int64
                                     10000 non-null
         Projects
                                                     int64
         Workshops/Certifications 10000 non-null
                                                     int64
         AptitudeTestScore 10000 non-null int64
SoftSkillsRating 10000 non-null float64
         SoftSkillsRating
         ExtracurricularActivities 10000 non-null object
     8 PlacementTraining 10000 non-null object
         SSC_Marks
                                    10000 non-null int64
     10 HSC_Marks
                                    10000 non-null int64
     11 PlacementStatus
                                    10000 non-null object
    dtypes: float64(2), int64(7), object(3)
    memory usage: 937.6+ KB
```

	Workshops/Cert	ifications	AptitudeTestScore	SoftSkillsRating
count	16	00000.00000	10000.000000	10000.000000
mean		1.013200	79.449900	4.323960
std		0.904272	8.159997	0.411622
min		0.000000	60.000000	3.000000
25%		0.000000	73.000000	4.000000
50%		1.000000	80.000000	4.400000
75%		2.000000	87.000000	4.700000
max		3.000000	90.000000	4.800000
	SSC_Marks	HSC_Marks	3	
count	10000.000000	10000.000000	)	
mean	69.159400	74.501500	)	
std	10.430459	8.919527	,	
min	55.000000	57.000000	)	
25%	59.000000	67.000000	)	
50%	70.000000	73.000000	)	
75%	78.000000	83.000000	)	
max	90,000000	88.000000	ነ	

Datase	t Description	:			
	StudentID	CGPA	Internships	Projects	\
count	10000.00000	10000.000000	10000.000000	10000.000000	
mean	5000.50000	7.698010	1.049200	2.026600	
std	2886.89568	0.640131	0.665901	0.867968	
min	1.00000	6.500000	0.000000	0.000000	
25%	2500.75000	7.400000	1.000000	1.000000	
50%	5000.50000	7.700000	1.000000	2.000000	
75%	7500.25000	8.200000	1.000000	3.000000	
max	10000.00000	9.100000	2.000000	3.000000	

F	irst 5 Rows	of the	Dataset:				
	StudentID	CGPA	Internships	Projects	Workshops/Certifi	cations.	1
Ø	1	7.5	1	1		1	
1	2	8.9	Ø	3		2	
2	3	7.3	1	2		2	
3	4	7.5	1	1		2	
4	5	8.3	1	2		2	
	AptitudeTe	estScore	softSkills	Rating Ext	racurricularActivi	ties \	
0		65	5	4.4		No	
1		96	)	4.0		Yes	
2		82	2	4.8		Yes	
3		85	5	4.4		Yes	
4		86	5	4.5		Yes	
	PlacementTi	raining	SSC_Marks	HSC_Marks	PlacementStatus		
0		No	61	79	NotPlaced		
1		Yes	78	82	Placed		
2		No	79	80	NotPlaced		
3		Yes	81	80	Placed		
4		Yes	74	88	Placed		
4		Yes	74	88	Placed		

# Drop columns that aren't useful.

```
0
```

```
columns_to_drop = ['StudentID']
df.drop(columns=columns_to_drop, inplace=True)
print(df.describe)
```

<box< th=""><th>method</th><th>NDFrame.describe</th><th>of</th><th>CGPA Internships Projects Workshops/Certifications</th><th>1</th></box<>	method	NDFrame.describe	of	CGPA Internships Projects Workshops/Certifications	1
Ø	7.5	1	1	1	
1	8.9	0	3	2	
2	7.3	1	2	2	
3	7.5	1	1	2	
4	8.3	1	2	2	
9995	7.5	1	1	2	
9996	7.4	0	1	0	
9997	8.4	1	3	0	
9998	8.9	0	3	2	
9999	8.4	Ø	1	1	

	AptitudeTestScore	SoftSkillsRating	Ext	tracurricularActivities	\				
Ø	65	4.4		No					
1	90	4.0	ı	Yes					
2	82	4.8		Yes					
3	85	4.4		Yes					
4	86	4.5		Yes					
9995	72	3.9		Yes					
9996	90	4.8		No					
9997	70	4.8		Yes					
9998	87	4.8		Yes					
9999	66	3.8		No					
	PlacementTraining	SSC_Marks HSC_Ma							
Ø	No	61	79	NotPlaced					
1	Yes	78	82	Placed					
2	No	79	80	NotPlaced					
3	Yes	81	80	Placed					
4	Yes	74	88	Placed					
• • •	•••	•••	• • •						
9995	No	85	66	NotPlaced					
9996	No	84	67	Placed					
9997	Yes	79	81	Placed					
9998	Yes	71	85	Placed					
9999	No	62	66	NotPlaced					
[1000	[10000 rows x 11 columns]>								

# Thus the columns now present in dataset are:

```
print("Dataset Information:")
    print(df.info())

→ Dataset Information:
    <class 'pandas.core.frame.DataFrame'>
    RangeIndex: 10000 entries, 0 to 9999
    Data columns (total 11 columns):
                                   Non-Null Count Dtype
    # Column
    Ø CGPA
                                   10000 non-null float64
        Internships
                                   10000 non-null int64
        Projects
                                  10000 non-null int64
        Workshops/Certifications 10000 non-null int64
AptitudeTestScore 10000 non-null int64
        AptitudeTestScore
        SoftSkillsRating
                                   10000 non-null float64
        ExtracurricularActivities 10000 non-null object
        PlacementTraining 10000 non-null object
        SSC Marks
                                   10000 non-null int64
     9 HSC_Marks
                                   10000 non-null int64
    10 PlacementStatus
                                  10000 non-null object
    dtypes: float64(2), int64(6), object(3)
    memory usage: 859.5+ KB
    None
```

## Take care of missing data.

c. Drop rows with maximum missing values.

```
print(f"Dataset Shape before Dropping Rows: {df.shape}")
# Drop rows with the highest number of missing values
threshold = len(df.columns) * 0.5 # Drop rows where over 50% of columns are missing
df = df.dropna(thresh=threshold)
print(f"Dataset Shape After Dropping Rows: {df.shape}")
Dataset Shape before Dropping Rows: (28671, 10)
Dataset Shape After Dropping Rows: (28671, 10)
```

[15]	<pre>print(df.isnull().sum())</pre>	
<b>→</b>	CGPA Internships Projects Workshops/Certifications AptitudeTestScore SoftSkillsRating ExtracurricularActivities PlacementTraining SSC_Marks HSC_Marks PlacementStatus dtype: int64	0 0 0 0 0 0 0 0

The output of print(df.isnull().sum()) indicates that there are no missing values in any of the columns of the dataset. Each column has a count of  $\theta$ , meaning that all records have valid data for every feature.

# **Data Preprocessing**

- The code uses dropna() to remove missing values.
- fillna() is used to fill missing values with the mean, specifically for numerical columns.
- get\_dummies() is applied to categorical variables like Placement Status, Placement Training, and Extracurricular Activities, with drop\_first=True to avoid multicollinearity.

```
[8] df.dropna(thresh=df.shape[1] - 1, inplace=True)
```

```
df.fillna(df.select_dtypes(include=['number']).mean(), inplace=True)
```

```
df_encoded = pd.get_dummies(df, columns=['PlacementStatus','PlacementTraining','ExtracurricularActivities'], drop_first=True)
print(df_encoded)
```

	CGPA	Internships	Projects	Worksho	ps/Certific	ations '	\
0	7.5	1	1			1	
1	8.9	0	3			2	
2	7.3	1	2			2	
3	7.5	1	1			2	
4	8.3	1	2			2	
9995	7.5	1	1			2	
9996	7.4	0	1			0	
9997	8.4	1	3			0	
9998	8.9	0	3			2	
9999	8.4	0	1			1	
	n		c - £+cl-: 11	-D-4:	SSC Marelan	UCC Marel	\
_	Артіт	udeTestScore	201.128111	_	SSC_Marks	HSC_Mark	-
0		65		4.4	61		79
1		90		4.0	78		32
2		82		4.8	79		30
3		85		4.4	81		30
4		86		4.5	74	i	38
		70			0.5		
9995		72		3.9	85		56
9996		90		4.8	84	(	57
9996 9997		90 <b>7</b> 0		4.8 4.8	84 <b>7</b> 9		57 31
9996		90		4.8	84	:	57

	PlacementStatus_Placed	PlacementTraining_Yes	A.	
0	False	False		
1	True	True		
2	False	False		
3	True	True		
4	True	True		
	•••	•••		
9995	False	False		
9996	True	False		
9997	True	True		
9998	True	True		
9999	False	False		
	ExtracurricularActiviti	es Yes		
0		False		
1		True		
2		True		
3		True		
4		True		
9995		True		
9996		False		
9997		True		
9998		True		
9999		False		
[1000	0 rows x 11 columns]			
LTOOO	O LOWS X II COTAIIII2			

### **Dataset Features**

- The dataset includes Aptitude Test Scores, Soft Skills Ratings, SSC Marks, HSC Marks, Internships, Projects, and Workshops/Certifications.
- It tracks whether a student is placed or not.

# **Saving Processed Data**

• The cleaned and encoded dataset is saved as "new\_data.csv".

```
[ ] df_encoded.to_csv("new_data.csv", index = False)

[ ] df_numeric = df.select_dtypes(include=[float, int])

Q1 = df_numeric.quantile(0.25)
Q3 = df_numeric.quantile(0.75)
IQR = Q3 - Q1

outliers = (df_numeric < (Q1 - 1.5 * IQR)) | (df_numeric > (Q3 + 1.5 * IQR))

df_outliers_marked = df.copy()

for col in df_numeric.columns:
    df_outliers_marked[col] = df[col].where(~outliers[col], other="OUTLIER")

print(df_outliers_marked)
```

82 Placed 80 NotPlaced 80 Placed 88 Placed 995 66 NotPlaced 996 67 Placed 997 81 Placed 998 85 Placed	HS	C_Marks	PlacementStatus	
80 NotPlaced 80 Placed 88 Placed 995 66 NotPlaced 996 67 Placed 997 81 Placed 998 85 Placed	)	79	NotPlaced	
80 Placed 88 Placed  995 66 NotPlaced 996 67 Placed 997 81 Placed 998 85 Placed	L	82	Placed	
88 Placed 995 66 NotPlaced 996 67 Placed 997 81 Placed 998 85 Placed	<u>)</u>	80	NotPlaced	
995 66 NotPlaced 996 67 Placed 997 81 Placed 998 85 Placed	}	80	Placed	
996 67 Placed 997 81 Placed 998 85 Placed	l	88	Placed	
996 67 Placed 997 81 Placed 998 85 Placed				
997 81 Placed 998 85 Placed	995	66	NotPlaced	
998 85 Placed	9996	67	Placed	
	9997	81	Placed	
999 66 NotPlaced	9998	85	Placed	
	999	66	NotPlaced	

# **Placement Analysis**

- There are binary indicators for Placement Status, Placement Training participation, and Extracurricular Activities.
- The dataset seems to have 10,000 rows and 11 columns.

### **Outlier Detection**

• Uses Interquartile Range (IQR) method to detect outliers in numerical data.

### **Saving Processed Data**

• The cleaned and encoded dataset is saved as "outliers marked.csv".

```
[ ] df_outliers_marked.to_csv("outliers_marked.csv", index=False)
[ ] scaler = StandardScaler()
[ ] df_numeric_scaled = pd.DataFrame(scaler.fit_transform(df_numeric), columns=df_numeric.columns)
      print("Standardized Dataframe:\n", df_numeric_scaled)

    Standardized Dataframe:

                   CGPA Internships Projects Workshops/Certifications \
          -0.309343 -0.073889 -1.182822
1.877818 -1.575689 1.121526
                                                                               1.091319
           -0.621794 -0.073889 -0.030648
-0.309343 -0.073889 -1.182822
                                                                               1.091319
                                                                               1.091319
             0.940464 -0.073889 -0.030648
                                                                               1.091319
     9995 -0.309343 -0.073889 -1.182822
9996 -0.465568 -1.575689 -1.182822
                                                                               1.091319
                                                                               -1.120516

    9997
    1.096689
    -0.073889
    1.121526

    9998
    1.877818
    -1.575689
    1.121526

    9999
    1.096689
    -1.575689
    -1.182822

                                                                               -1.120516
                                                                               1.091319
                                                                               -0.014598
```

	AptitudeTestScore	SoftSkillsRating	SSC_Marks	HSC_Marks
0	-1.770910	0.184742	-0.782306	0.504368
1	1.292970	-0.787072	0.847618	0.840726
2	0.312528	1.156555	0.943496	0.616487
3	0.680194	0.184742	1.135251	0.616487
4	0.802749	0.427695	0.464106	1.513441
9995	-0.913024	-1.030025	1.518763	-0.953181
9996	1.292970	1.156555	1.422885	-0.841062
9997	-1.158134	1.156555	0.943496	0.728606
9998	0.925304	1.156555	0.176473	1.177083
9999	-1.648355	-1.272978	-0.686428	-0.953181
[1000	00 rows x 8 columns]			

## 1. Normalization Technique:

- MinMaxScaler() is applied to scale all numerical features between 0 and 1.
- This transformation ensures that all values lie within a uniform range, preventing dominance by larger values.

#### 2. Affected Features:

The normalized dataset includes:

- CGPA
- Internships
- Projects
- Workshops/Certifications
- Aptitude Test Score
- Soft Skills Rating
- o SSC Marks
- HSC Marks

## 3. Purpose:

- Ensures that all numerical features contribute equally to the analysis/model.
- Helps in improving the performance of machine learning algorithms, particularly those that rely on distance-based calculations (e.g., KNN, SVM, Neural Networks).

#### What It Indicates

- The dataset is being prepared for machine learning by applying feature scaling.
- The placement prediction model will likely use this normalized data to enhance accuracy and avoid bias due to large variations in feature magnitudes.

```
normalizer = MinMaxScaler()
       df_numeric_normalized = pd.DataFrame(normalizer.fit_transform(df_numeric_scaled), columns=df_numeric.columns)
       print("Standardized and Normalized Dataframe:\n", df_numeric_normalized)

→ Standardized and Normalized Dataframe:

                        CGPA Internships Projects Workshops/Certifications \
           0.333333
                                                                                                    0.666667
                                                                                                   0.666667
      0.666667
                                                                                                 0.666667
                                                                                              0.666667
0.000000
                                                                                                    0.000000
                                                                                                    0.666667
                                                                                                    0.333333
                AptitudeTestScore SoftSkillsRating SSC_Marks HSC_Marks
                    0.166667 0.777778 0.171429 0.709677

      0.166667
      0.777778
      0.171429
      0.709677

      1.000000
      0.555556
      0.657143
      0.806452

      0.733333
      1.000000
      0.685714
      0.741935

      0.833333
      0.777778
      0.742857
      0.741935

      0.866667
      0.833333
      0.542857
      1.000000

      1.000000
      0.857143
      0.290323

      1.000000
      0.825714
      0.322581

      0.333333
      1.000000
      0.685714
      0.774194

      0.90000
      1.000000
      0.457143
      0.903226

      0.200000
      0.444444
      0.200000
      0.290323

      1
       2
       4
       9995
       9996
       9997
                                                            1.000000 0.457143 0.903226
0.444444 0.200000 0.290323
       [10000 rows \times 8 columns]
```

Conclusion: In conclusion, this experiment highlighted the significance of proper data cleaning and preparation to ensure high-quality data. We addressed key issues such as missing values, irrelevant data, and outliers that could negatively impact the dataset and ultimately distort model performance. Missing values were managed through imputation or removal methods, irrelevant features were discarded, and outliers were carefully handled to avoid skewing the results. These steps were crucial for creating a dataset that is accurate and reliable for analysis.

Additionally, the dataset was scaled for uniformity, which is important to ensure that all features are treated equally. By applying standardization or normalization techniques, we ensured that the model could work efficiently without being influenced by differences in scale between variables. Overall, these data preparation techniques are fundamental in setting up a clean, well-structured dataset, which is essential for generating reliable model outcomes and drawing meaningful conclusions.