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In [ ]: #slip20
# Q1) Read "StudentsPerformance.csv" and display dataset information

import pandas as pd

# Read the CSV file
data = pd.read_csv("StudentsPerformance.csv")

# Display the shape of the dataset (rows, columns)
print("Shape of dataset (rows, columns):", data.shape)

# Display the first 5 rows of the dataset
print("\nTop rows of the dataset:")
print(data.head())

# Display column names
print("\nColumns in dataset:")
print(data.columns.tolist())

```

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In [3]: #slip19
import pandas as pd
import matplotlib.pyplot as plt

# create sample dataset (if CSV given, use pd.read_csv("Seller.csv"))
data = pd.DataFrame({
    "SellerType": ["Individual", "Company", "Individual", "Company", "Agency",
                  "Agency", "Individual", "Company", "Agency", "Company"],
    "LayoutType": ["2BHK", "3BHK", "1BHK", "2BHK", "3BHK",
                  "2BHK", "1BHK", "2BHK", "3BHK", "2BHK"]
})

# a) Count different seller types
print("Different seller types:\n", data["SellerType"].value_counts())
data["SellerType"].value_counts().plot(kind="bar", title="Seller Types")
plt.show()

# b) Seller with minimum records
print("\nSeller with minimum records:\n", data["SellerType"].value_counts().idxmin())

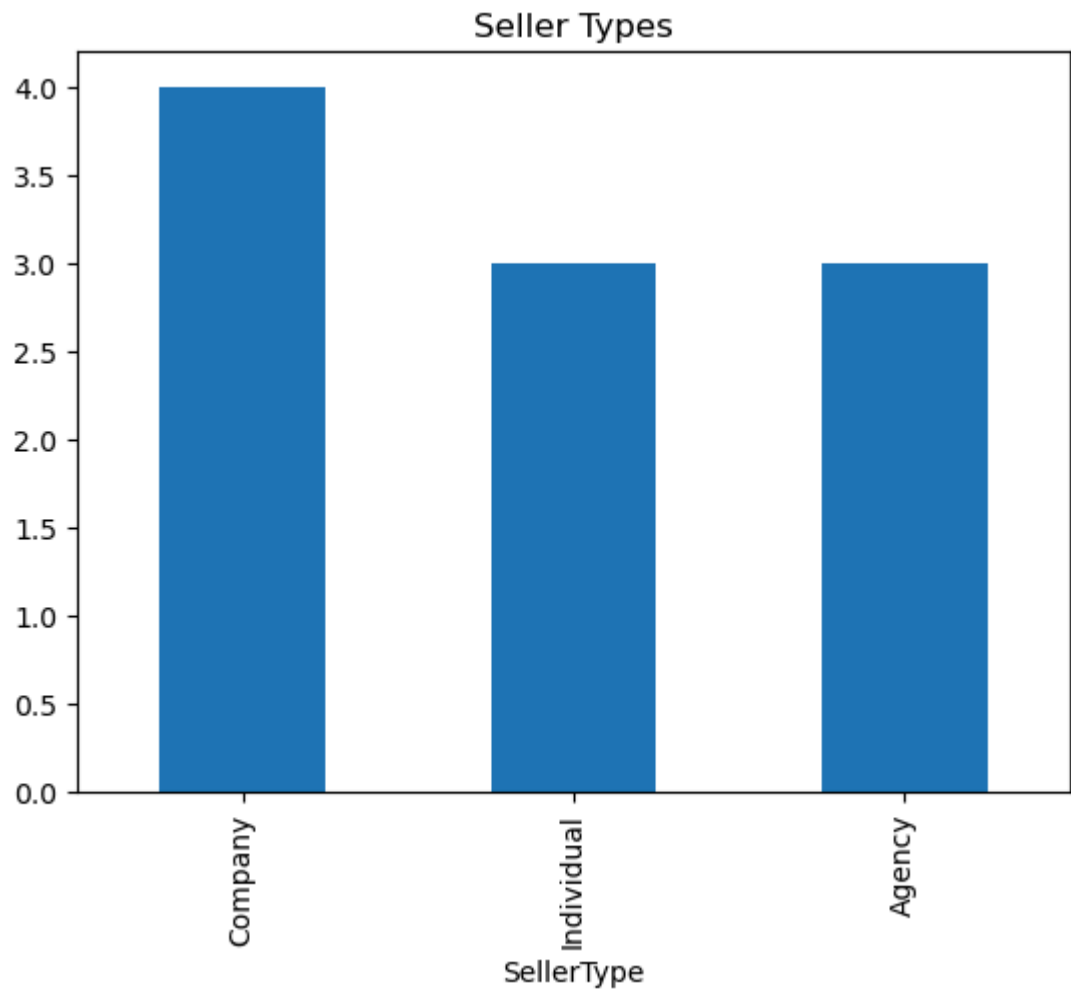
# c) Count of layout types
print("\nDifferent layout types:\n", data["LayoutType"].value_counts())
data["LayoutType"].value_counts().plot(kind="bar", title="Layout Types")
plt.show()

```

Different seller types:

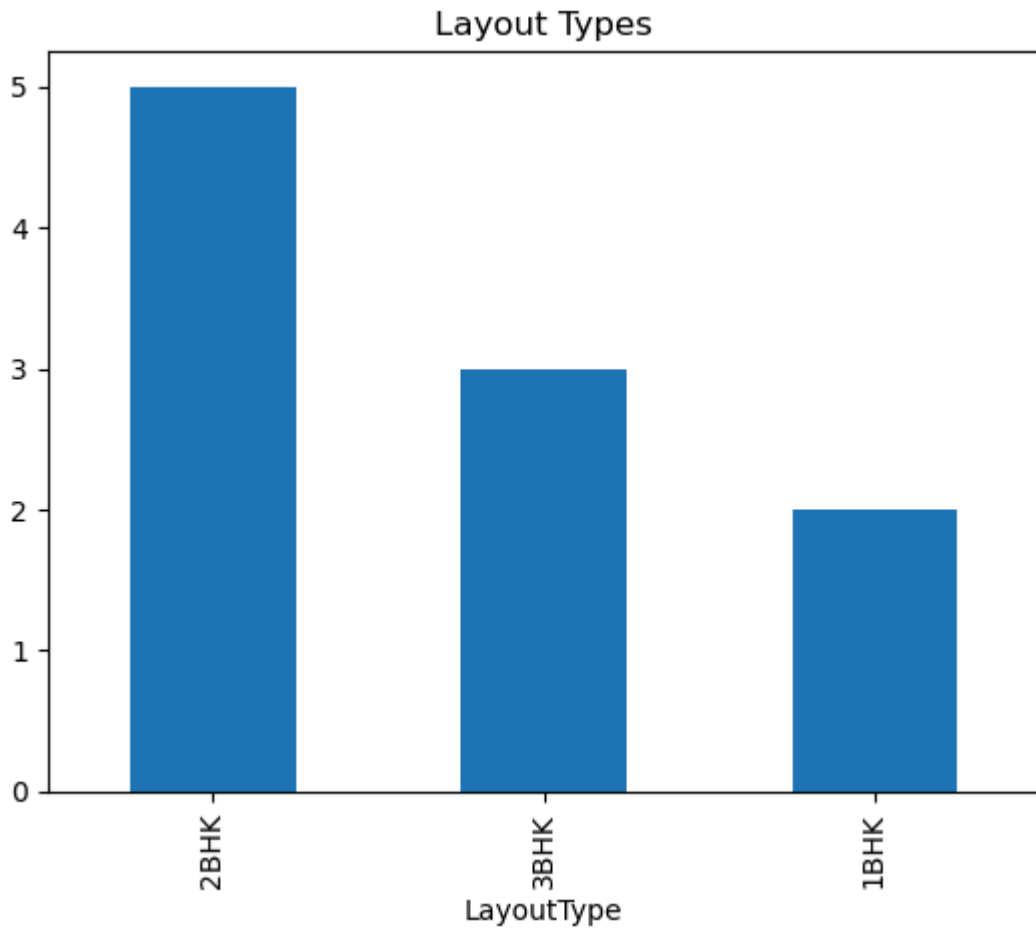
SellerType	count
Company	4
Individual	3
Agency	3

Name: count, dtype: int64



Seller with minimum records:
Individual

Different layout types:
LayoutType
2BHK 5
3BHK 3
1BHK 2
Name: count, dtype: int64



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In [4]: #slip 16
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns

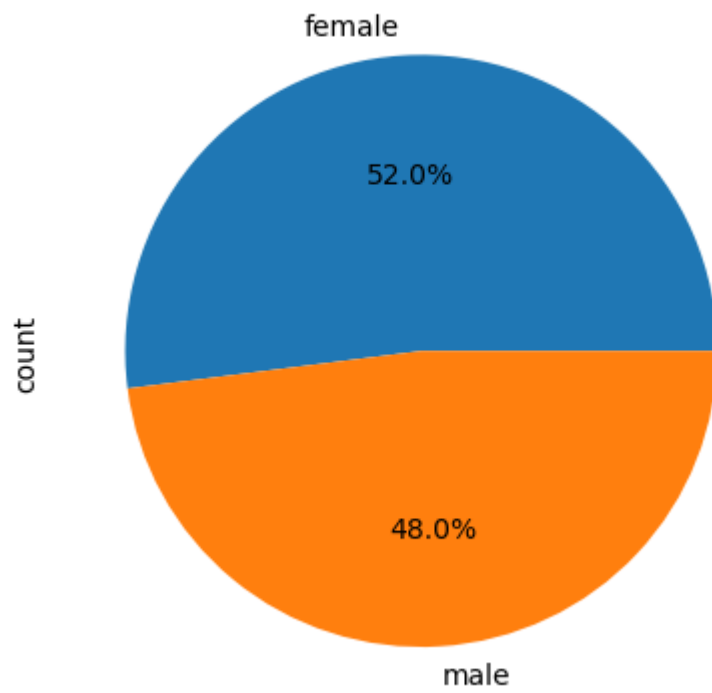
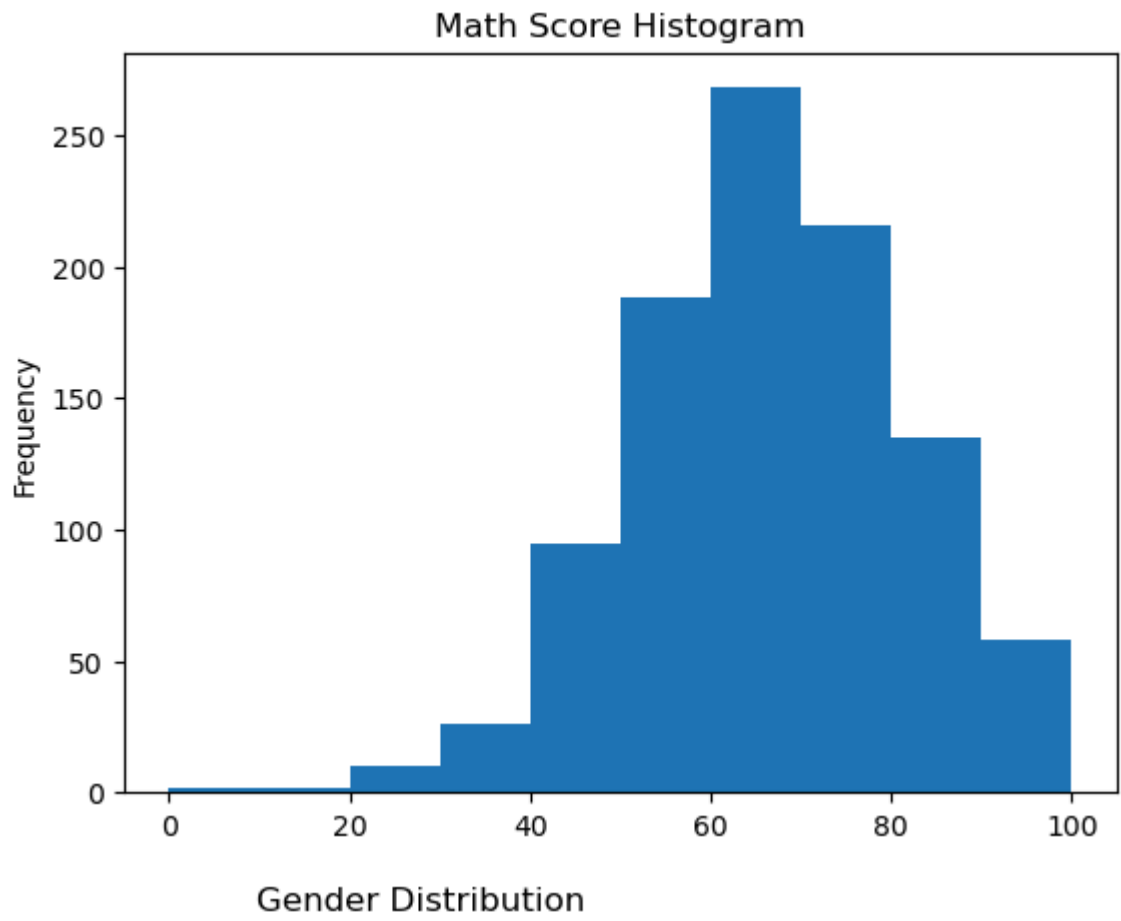
# Read CSV (or create dummy data)
data = pd.read_csv("Student.csv")

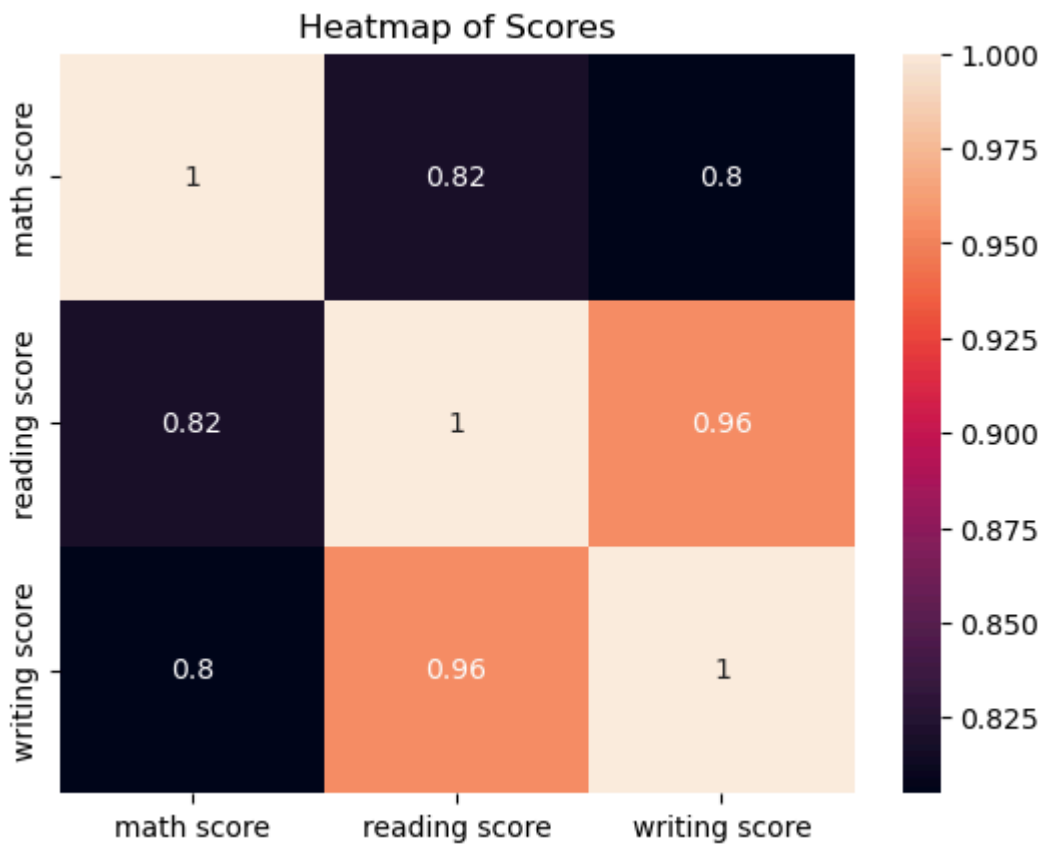
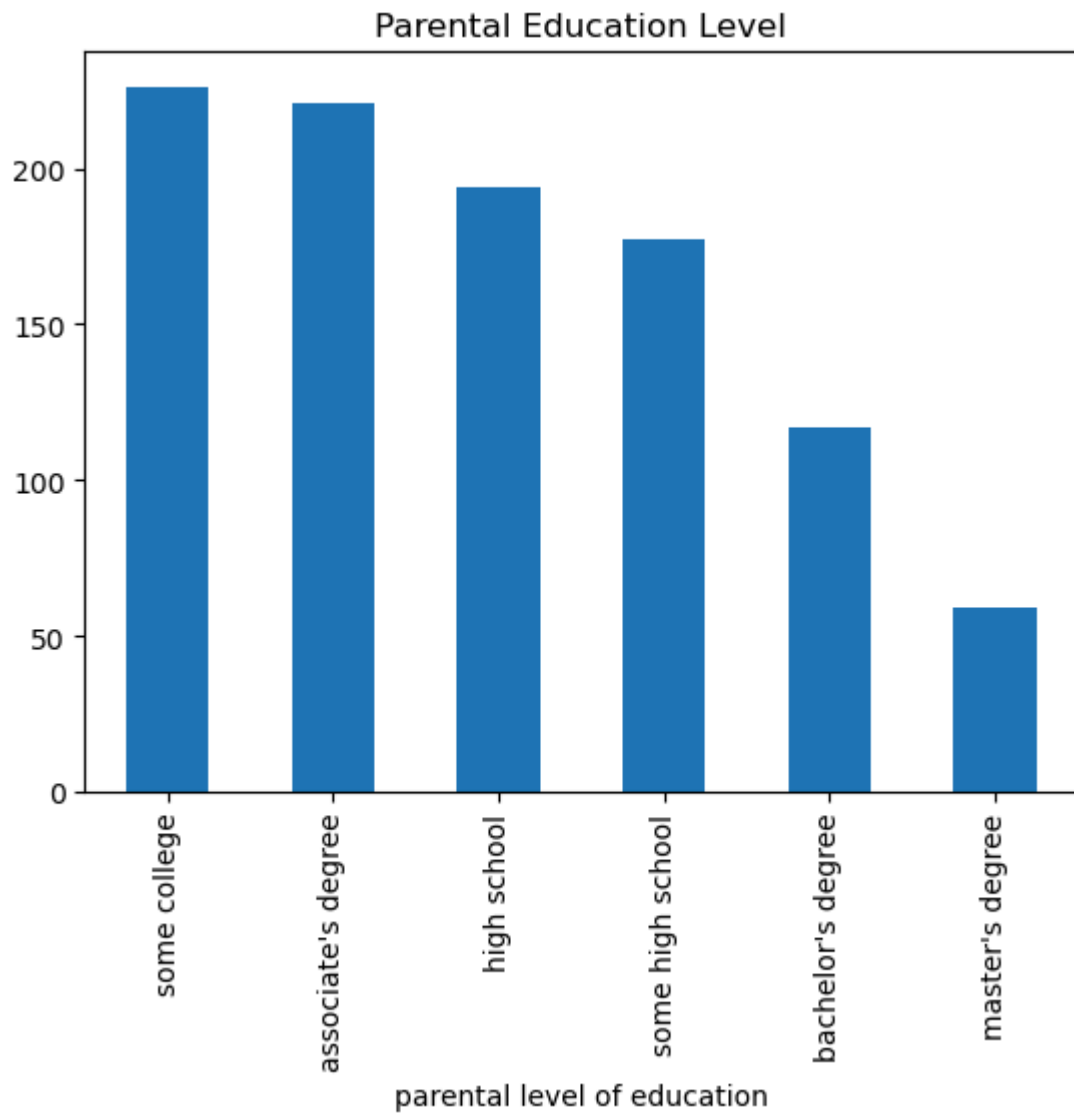
# 1) Histogram of Math Score
data['math score'].plot(kind='hist', title='Math Score Histogram', bins=10)
plt.show()

# 2) Pie Chart of Gender
data['gender'].value_counts().plot(kind='pie', autopct='%1.1f%%', title='Gender')
plt.show()

# 3) Bar Graph of Parental Level of Education
data['parental level of education'].value_counts().plot(kind='bar', title='Parental Level of Education')
plt.show()

# 4) Heatmap of correlation between scores
sns.heatmap(data[['math score', 'reading score', 'writing score']].corr(), annot=True)
plt.title("Heatmap of Scores")
plt.show()
```





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In [5]: #slip17
import numpy as np

# 1) np.array()
arr1D = np.array([1, 2, 3, 4, 5])      # 1D array
arr2D = np.array([[1, 2], [3, 4]])    # 2D array
print("1D array:\n", arr1D)
print("2D array:\n", arr2D)

# 2) np.arange()
arr_range = np.arange(0, 10, 2)        # 0 to 10 step 2
print("\nArange array:\n", arr_range)

# 3) np.zeros()
zeros1D = np.zeros(5)                  # 1D zeros
zeros2D = np.zeros((2,3))              # 2D zeros
print("\n1D zeros:\n", zeros1D)
print("2D zeros:\n", zeros2D)

# 4) np.ones()
ones1D = np.ones(5)                    # 1D ones
ones2D = np.ones((2,3))                # 2D ones
print("\n1D ones:\n", ones1D)
print("2D ones:\n", ones2D)

```

1D array:
[1 2 3 4 5]

2D array:
[[1 2]
[3 4]]

Arange array:
[0 2 4 6 8]

1D zeros:
[0. 0. 0. 0. 0.]

2D zeros:
[[0. 0. 0.]
[0. 0. 0.]]

1D ones:
[1. 1. 1. 1. 1.]

2D ones:
[[1. 1. 1.]
[1. 1. 1.]]

```

In [6]: #slip14
import numpy as np

# Create a sample NumPy array
arr = np.array([10, 20, 30, 40, 50])

# 1) Sum
print("Sum:", np.sum(arr))

# 2) Mean
print("Mean:", np.mean(arr))

# 3) Median
print("Median:", np.median(arr))

```

```

# 4) Variance
print("Variance:", np.var(arr))

# 5) Standard Deviation
print("Std Dev:", np.std(arr))

# 6) Minimum
print("Min:", np.min(arr))

# 7) Maximum
print("Max:", np.max(arr))

```

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Sum: 150
Mean: 30.0
Median: 30.0
Variance: 200.0
Std Dev: 14.142135623730951
Min: 10
Max: 50

```

```

In [7]: #slip13
import pandas as pd
import numpy as np

# 1D Array → DataFrame
arr1D = np.array([10, 20, 30, 40, 50])
df1D = pd.DataFrame(arr1D, columns=["Numbers"])
print("1D DataFrame:\n", df1D)

# Indexing 1D
print("\n1D Indexing (3rd element):", df1D.iloc[2,0])

# Slicing 1D
print("1D Slicing (2nd to 4th element):\n", df1D.iloc[1:4, 0])

# 2D Array → DataFrame
arr2D = np.array([[1,2,3],[4,5,6],[7,8,9]])
df2D = pd.DataFrame(arr2D, columns=["A","B","C"])
print("\n2D DataFrame:\n", df2D)

# Indexing 2D
print("\n2D Indexing (row 2, column B):", df2D.iloc[1,1])

# Slicing 2D
print("2D Slicing (first 2 rows, first 2 columns):\n", df2D.iloc[0:2, 0:2])

```

1D DataFrame:

```
Numbers
0      10
1      20
2      30
3      40
4      50
```

1D Indexing (3rd element): 30

1D Slicing (2nd to 4th element):

```
1      20
2      30
3      40
```

Name: Numbers, dtype: int64

2D DataFrame:

```
   A  B  C
0  1  2  3
1  4  5  6
2  7  8  9
```

2D Indexing (row 2, column B): 5

2D Slicing (first 2 rows, first 2 columns):

```
   A  B
0  1  2
1  4  5
```

```
In [8]: #slip11
import pandas as pd

# Create Employee DataFrame with 7 columns
data = {
    "EmpID": [101, 102, 103, 104, 105],
    "Name": ["Amit", "Sneha", "Ravi", "Priya", "Karan"],
    "Age": [25, 28, 24, 27, 30],
    "Department": ["HR", "IT", "Finance", "IT", "Marketing"],
    "Salary": [50000, 60000, 55000, 58000, 62000],
    "Experience": [2, 5, 1, 4, 6],
    "City": ["Mumbai", "Delhi", "Pune", "Bangalore", "Hyderabad"]
}

df = pd.DataFrame(data)

# Display the DataFrame
print(df)
```

	EmpID	Name	Age	Department	Salary	Experience	City
0	101	Amit	25	HR	50000	2	Mumbai
1	102	Sneha	28	IT	60000	5	Delhi
2	103	Ravi	24	Finance	55000	1	Pune
3	104	Priya	27	IT	58000	4	Bangalore
4	105	Karan	30	Marketing	62000	6	Hyderabad

```
In [9]: #slip10
import numpy as np

# Create a 2D array
arr = np.array([[1, 2, 3, 4], [5, 6, 7, 8]])
print("Original Array:\n", arr)
```



```

# 1) Reshape (change shape without changing data)
reshaped = arr.reshape(4, 2) # 4 rows, 2 columns
print("\nReshaped Array (4x2):\n", reshaped)

# 2) Resize (change shape and can change total size)
resized = np.resize(arr, (3, 3)) # 3x3 array
print("\nResized Array (3x3):\n", resized)

# 3) Transpose (swap rows and columns)
transposed = arr.T
print("\nTransposed Array:\n", transposed)

```

Original Array:

```

[[1 2 3 4]
 [5 6 7 8]]

```

Reshaped Array (4x2):

```

[[1 2]
 [3 4]
 [5 6]
 [7 8]]

```

Resized Array (3x3):

```

[[1 2 3]
 [4 5 6]
 [7 8 1]]

```

Transposed Array:

```

[[1 5]
 [2 6]
 [3 7]
 [4 8]]

```

```

In [10]: #slip9
import numpy as np

# Create a 2D array
arr = np.array([[1, 2, 3], [4, 5, 6]])
print("\n2D Array:\n", arr)

# Shape of the array (rows, columns)
print("\nShape:", arr.shape)

# Total number of elements
print("\nSize:", arr.size)

# Number of dimensions
print("\nDimensions:", arr.ndim)

```

2D Array:

```

[[1 2 3]
 [4 5 6]]

```

Shape: (2, 3)

Size: 6

Dimensions: 2

```

In [11]: #slip8
import pandas as pd

# Create sample dataset

```

```

data = pd.DataFrame({
    "Student": ["Amit", "Sneha", "Ravi", "Priya", "Karan"],
    "Marks": [32, 28, 35, 40, 30],
    "Attempts": [1, 2, 1, 3, 1]
})

print("Dataset:\n", data)

# a) Rows where marks are between 30 and 35
marks_30_35 = data[(data["Marks"] >= 30) & (data["Marks"] <= 35)]
print("\nMarks between 30 and 35:\n", marks_30_35)

# b) Rows where attempts < 2 and marks > 30
attempts_marks = data[(data["Attempts"] < 2) & (data["Marks"] > 30)]
print("\nAttempts < 2 and Marks > 30:\n", attempts_marks)

# c) Sum of examination attempts
total_attempts = data["Attempts"].sum()
print("\nTotal examination attempts:", total_attempts)

```

Dataset:

	Student	Marks	Attempts
0	Amit	32	1
1	Sneha	28	2
2	Ravi	35	1
3	Priya	40	3
4	Karan	30	1

Marks between 30 and 35:

	Student	Marks	Attempts
0	Amit	32	1
2	Ravi	35	1
4	Karan	30	1

Attempts < 2 and Marks > 30:

	Student	Marks	Attempts
0	Amit	32	1
2	Ravi	35	1

Total examination attempts: 8

```

In [12]: #slip7
import pandas as pd

# Create sample DataFrame
df = pd.DataFrame({
    "Math": [50, 60, 70],
    "Science": [55, 65, 75]
})

print("Original DataFrame:\n", df)

# a) apply() → apply function to each column
print("\nUsing apply() to add 5 to each column:")
print(df.apply(lambda x: x + 5))

# b) applymap() → apply function to each element
print("\nUsing applymap() to add 10 to each element:")
print(df.applymap(lambda x: x + 10))

```

```
# c) map() → apply function to a single column
print("\nUsing map() on Math column to double the values:")
print(df["Math"].map(lambda x: x * 2))
```

Original DataFrame:

	Math	Science
0	50	55
1	60	65
2	70	75

Using apply() to add 5 to each column:

	Math	Science
0	55	60
1	65	70
2	75	80

Using applymap() to add 10 to each element:

	Math	Science
0	60	65
1	70	75
2	80	85

Using map() on Math column to double the values:

0	100
1	120
2	140

Name: Math, dtype: int64

C:\Users\User\AppData\Local\Temp\ipykernel_29848\2794539646.py:18: FutureWarning: DataFrame.applymap has been deprecated. Use DataFrame.map instead.
 print(df.applymap(lambda x: x + 10))

```
In [13]: #slip4
import numpy as np
from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_squared_error, r2_score

# Data
x = np.array([1,2,3,4,5,6,7,8]).reshape(-1,1) # x must be 2D
y = np.array([7,14,15,18,19,21,26,23])

# Create Linear Regression model
model = LinearRegression()
model.fit(x, y)

# Estimated coefficients
m = model.coef_[0] # slope
c = model.intercept_ # intercept
print("Estimated slope (m):", m)
print("Estimated intercept (c):", c)

# Predictions
y_pred = model.predict(x)

# Performance metrics
mse = mean_squared_error(y, y_pred)
r2 = r2_score(y, y_pred)
print("\nMean Squared Error (MSE):", mse)
print("R-squared (R2 score):", r2)
```

Estimated slope (m): 2.2738095238095237
Estimated intercept (c): 7.642857142857142

Mean Squared Error (MSE): 3.4657738095238084
R-squared (R2 score): 0.8867741072947811

```
In [14]: #slip3
import pandas as pd

# Create a dictionary
data = {
    "EmpID": [101, 102, 103],
    "Name": ["Amit", "Sneha", "Ravi"],
    "Department": ["HR", "IT", "Finance"],
    "Salary": [50000, 60000, 55000]
}

# Convert dictionary to DataFrame
df = pd.DataFrame(data)

# Display the DataFrame
print("Employee Data:\n", df)
```

Employee Data:

	EmpID	Name	Department	Salary
0	101	Amit	HR	50000
1	102	Sneha	IT	60000
2	103	Ravi	Finance	55000

```
In [15]: #slip1
import pandas as pd

# Create sample dictionary
data = {
    "Employee": ["Amit", "Sneha", "Ravi", "Amit", "Ravi"],
    "Department": ["HR", "IT", "Finance", "HR", "Finance"],
    "Salary": [50000, 60000, 55000, 52000, 58000]
}

# Convert dictionary to DataFrame
df = pd.DataFrame(data)

# Create Pivot Table: sum of Salary by Employee and Department
pivot = pd.pivot_table(df, index="Employee", columns="Department", values="Salary")

print("Pivot Table:\n", pivot)
```

Pivot Table:

Employee	Finance	HR	IT
Amit	0	102000	0
Ravi	113000	0	0
Sneha	0	0	60000

In []: