Department of computer science

Foundation of Data Science-BCA III

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Program 1: Perform Arithmetic Operations on NumPy Arrays

```
Code: import numpy as np
a = np.array([1, 2, 3])
b = np.array([4, 5, 6])
print("Sum:", a + b)
print("Product:", a * b)
```

Output:

Sum: [5 7 9]

Product: [4 10 18]

Program 2: Filter Rows Based on Condition in Pandas

```
import pandas as pd
df = pd.DataFrame({'Name': ['A', 'B', 'C'], 'Marks': [45, 80, 65]})
print("Marks > 60:\n", df[df['Marks'] > 60])
Output:
Marks > 60:
 Name Marks
1 B
        80
2 C
        65
Program 3: Frequency Count using Value Counts in Pandas
Code: import pandas as pd
df = pd.DataFrame({'Fruit': ['Apple', 'Banana', 'Apple', 'Orange', 'Banana']})
print(df['Fruit'].value counts())
Output:
Fruit
Apple
         2
Banana
         2
Orange
        1
Name: count, dtype: int64
Program 4: Write a python program for
  a) Conditional statements
  b) Loops
  c) Functions
Code:
Conditional statements:
num = int(input("Enter a number: "))
if num > 0:
  print("The number is Positive")
elif num < 0:
  print("The number is Negative")
else:
  print("The number is Zero")
Output:
Enter a number: -7
The number is Negative
```

```
Loops:
for i in range(1, 11):
  print(i)
Output:
1
2
3
4
5
6
7
8
9
10
Functions:
def square(num):
  return num * num
n = int(input("Enter a number: "))
print("Square of", n, "is", square(n))
Output:
Enter a number: 5
Square of 5 is 25
Program 5: Data Loading, Storage
Code:
import pandas as pd
from io import StringIO
data = {'Name': ['Ali', 'Sara', 'John'], 'Marks': [85, 90, 78]}
df = pd.DataFrame(data)
# Simulate CSV file using StringIO
csv buffer = StringIO()
df.to csv(csv buffer, index=False)
csv buffer.seek(0)
# Load data from the simulated CSV
```

```
df loaded = pd.read csv(csv buffer)
print("Original Data:")
print(df)
print("\nLoaded Data from CSV:")
print(df loaded)
Output:
Original Data:
 Name Marks
0 Ali
        85
1 Sara
         90
2 John
         78
Loaded Data from CSV:
 Name Marks
0 Ali
        85
1 Sara
         90
         78
2 John
Program 6: Data Cleaning and Preparation.
Code:
import pandas as pd
data = {
  'Name': ['Ali', 'Sara', None, 'John', 'Ali'],
  'Marks': [85, 90, None, 78, 150],
  'City': ['Delhi', 'Mumbai', 'Delhi', None, 'Delhi']
}
df = pd.DataFrame(data)
print("Original Data:")
print(df)
# Remove duplicates
df = df.drop duplicates()
# Handle missing values
df = df.fillna({'Name': 'Unknown', 'Marks': df['Marks'].mean(), 'City':
'Unknown'})
# Rename columns
```

```
df = df.rename(columns={'Name': 'Student Name', 'Marks': 'Score', 'City':
'Location'})
# Remove outliers (Marks > 100)
df = df[df['Score'] <= 100]
# Convert names to title case
df['Student Name'] = df['Student Name'].str.title()
print("\nCleaned & Prepared Data:")
print(df)
Output:
Original Data:
  Name Marks City
0 Ali 85.0 Delhi
1 Sara 90.0 Mumbai
2 None NaN Delhi
3 John 78.0 None
4 Ali 150.0 Delhi
Cleaned & Prepared Data:
 Student Name Score Location
      Ali 85.0 Delhi
0
     Sara 90.0 Mumbai
1
     John 78.0 Unknown
3
    Unknown 84.3 Delhi
2
Program 7: Data Manipulation with Pandas.
Code:
import pandas as pd
data = {
  'Name': ['Ali', 'Sara', 'John', 'Emma'],
  'Marks': [85, 90, 78, 92],
  'City': ['Delhi', 'Mumbai', 'Delhi', 'Chennai']
```

df = pd.DataFrame(data)

```
print("Original Data:")
print(df)
# Select specific columns
print("\nSelect Name and Marks:")
print(df[['Name', 'Marks']])
# Filter rows where Marks > 80
print("\nStudents with Marks > 80:")
print(df[df['Marks'] > 80])
# Sort by Marks descending
print("\nSorted by Marks (Descending):")
print(df.sort values(by='Marks', ascending=False))
# Add a new column (Grade)
df['Grade'] = ['B', 'A', 'C', 'A']
print("\nAfter Adding Grade Column:")
print(df)
# Group by City and calculate average Marks
print("\nAverage Marks by City:")
print(df.groupby('City')['Marks'].mean())
Output:
Original Data:
  Name Marks City
0 Ali
        85 Delhi
1 Sara 90 Mumbai
2 John 78 Delhi
           92 Chennai
3 Emma
Select Name and Marks:
  Name Marks
0 Ali
        85
1 Sara 90
2 John
         78
3 Emma
           92
```

Students with Marks > 80:

Name Marks City

- 0 Ali 85 Delhi
- 1 Sara 90 Mumbai
- 3 Emma 92 Chennai

Sorted by Marks (Descending):

Name Marks City

- 3 Emma 92 Chennai
- 1 Sara 90 Mumbai
- 0 Ali 85 Delhi
- 2 John 78 Delhi

After Adding Grade Column:

Name Marks City Grade

- 0 Ali 85 Delhi B
- 1 Sara 90 Mumbai A
- 2 John 78 Delhi C
- 3 Emma 92 Chennai A

Average Marks by City:

City

Chennai 92.0

Delhi 81.5

Mumbai 90.0

Name: Marks, dtype: float64

Program 8: Plotting and Visualization.

Code:

import pandas as pd

import matplotlib.pyplot as plt

data = {'Name': ['Ali', 'Sara', 'John', 'Emma'],

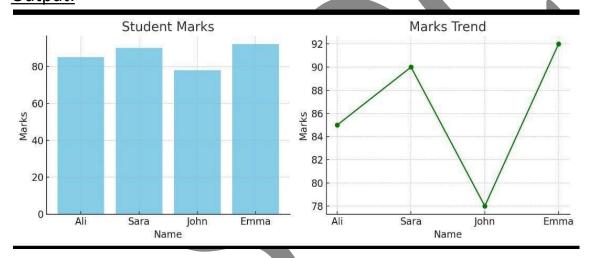
'Marks': [85, 90, 78, 92]}

df = pd.DataFrame(data)

Bar Plot

plt.figure(figsize=(6,4))

```
plt.bar(df['Name'], df['Marks'], color='skyblue')
plt.title('Student Marks')
plt.xlabel('Name')
plt.ylabel('Marks')
plt.show()
# Line Plot
plt.figure(figsize=(6,4))
plt.plot(df['Name'], df['Marks'], marker='o', color='green')
plt.title('Marks Trend')
plt.xlabel('Name')
plt.ylabel('Name')
plt.show()
Output:
```



Program 9: Advanced Numpy.

Code:

import numpy as np

1. Create a 2D array

data = np.array([[10, 20, 30], [40, 50, 60], [70, 80, 90]])

print("Original Array:\n", data)

2. Broadcasting (Add 5 to every element)

broadcast_result = data + 5

print("\nAfter Broadcasting (Add 5):\n", broadcast_result)

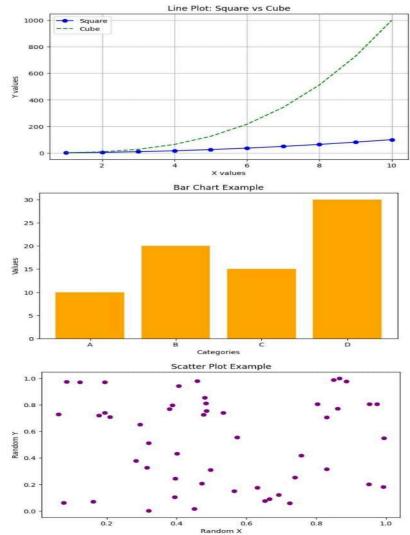
3. Boolean Indexing (Find values > 50)

```
greater than 50 = data[data > 50]
print("\nValues greater than 50:\n", greater_than_50)
# 4. Fancy Indexing (Select specific rows and columns)
fancy result = data[[0, 2], [1, 2]] # (row0,col1) and (row2,col2)
print("\nFancy Indexing Result:\n", fancy result)
# 5. Vectorized Operation (Square of all elements)
squared = np.square(data)
print("\nSquared Elements:\n", squared)
# 6. Aggregate Functions (mean, sum, std)
print("\nMean:", np.mean(data))
print("Sum:", np.sum(data))
print("Standard Deviation:", np.std(data))
#7. Reshape (Convert 3x3 to 1x9)
reshaped = data.reshape(1, 9)
print("\nReshaped Array (1x9):\n", reshaped)
Output:
Original Array:
[[10 20 30]
[40 50 60]
[70 80 90]]
After Broadcasting (Add 5):
[[15 25 35]
[45 55 65]
[75 85 95]]
Values greater than 50:
[60 70 80 90]
Fancy Indexing Result:
[20 90]
Squared Elements:
[[ 100 400 900]
[1600 2500 3600]
[4900 6400 8100]]
Mean: 50.0
Sum: 450
```

```
Standard Deviation: 25.81988897471611
Reshaped Array (1x9):
[[10 20 30 40 50 60 70 80 90]]
Program 10: Matplotlib
Code:
import matplotlib.pyplot as plt
import numpy as np
# Sample Data
x = np.arange(1, 11)
y1 = x ** 2
y2 = x ** 3
# 1. Line Plot
plt.figure(figsize=(8, 5))
plt.plot(x, y1, label='Square', color='blue', marker='o')
plt.plot(x, y2, label='Cube', color='green', linestyle='--')
plt.title("Line Plot: Square vs Cube")
plt.xlabel("X values")
plt.ylabel("Y values")
plt.legend()
plt.grid(True)
plt.show()
#2. Bar Chart
plt.figure(figsize=(8, 5))
categories = ['A', 'B', 'C', 'D']
values = [10, 20, 15, 30]
plt.bar(categories, values, color='orange')
plt.title("Bar Chart Example")
plt.xlabel("Categories")
plt.ylabel("Values")
plt.show()
#3. Scatter Plot
plt.figure(figsize=(8, 5))
x scatter = np.random.rand(50)
y scatter = np.random.rand(50)
```

plt.scatter(x_scatter, y_scatter, color='purple')
plt.title("Scatter Plot Example")
plt.xlabel("Random X")
plt.ylabel("Random Y")
plt.show()

Output:



Program 11: Building and optimizing pipelines in scikit-learn.

Code:

from sklearn.datasets import load_iris from sklearn.model_selection import train_test_split, GridSearchCV from sklearn.preprocessing import StandardScaler from sklearn.linear_model import LogisticRegression

```
# 1. Load dataset
iris = load iris()
X, y = iris.data, iris.target
# 2. Split data into train and test
X train, X test, y train, y test = train test split(X, y, test size=0.3,
random state=42)
# 3. Create a pipeline: Scaling + Logistic Regression
pipeline = Pipeline([
  ('scaler', StandardScaler()), # Feature scaling
  ('model', LogisticRegression(max_iter=200)) # Model
])
# 4. Define hyperparameter grid for optimization
param grid = {
                                    # Regularization strength
  'model C': [0.01, 0.1, 1, 10],
  'model__solver': ['lbfgs', 'liblinear'] # Solver options
}
# 5. Use GridSearchCV for optimization
grid_search = GridSearchCV(pipeline, param_grid, cv=5, scoring='accuracy')
grid_search.fit(X_train, y_train)
# 6. Print best parameters and scores
print("Best Parameters:", grid search.best params )
print("Best CV Score:", grid search.best score )
# 7. Test set accuracy
test accuracy = grid search.score(X test, y test)
print("Test Accuracy:", test accuracy)
Output:
```

Best Parameters: {'model_C': 1, 'model_solver': 'lbfgs'}

Best CV Score: 0.975 Test Accuracy: 0.978

Program 12: BOX PLOT, Write a Python program using Matplotlib to create a boxplot that shows the distribution of marks for students in three subjects: Math, Science, and English. Add appropriate labels, a title, and display the plot.

Code:

```
import matplotlib.pyplot as plt

# Sample data: Marks of students in three subjects
math = [85, 90, 78, 92, 88, 76, 95]
science = [80, 85, 79, 91, 84, 77, 89]
english = [78, 82, 85, 88, 90, 83, 87]

# Combine all subjects
data = [math, science, english]

# Create boxplot
plt.figure(figsize=(8, 5))
plt.boxplot(data, labels=['Math', 'Science', 'English'])
plt.title("Boxplot of Student Marks")
plt.ylabel("Marks")
plt.grid(True)
plt.show()
```

Output:

