DATE : 21/03/2023

ASSIGNMENT NO – 11

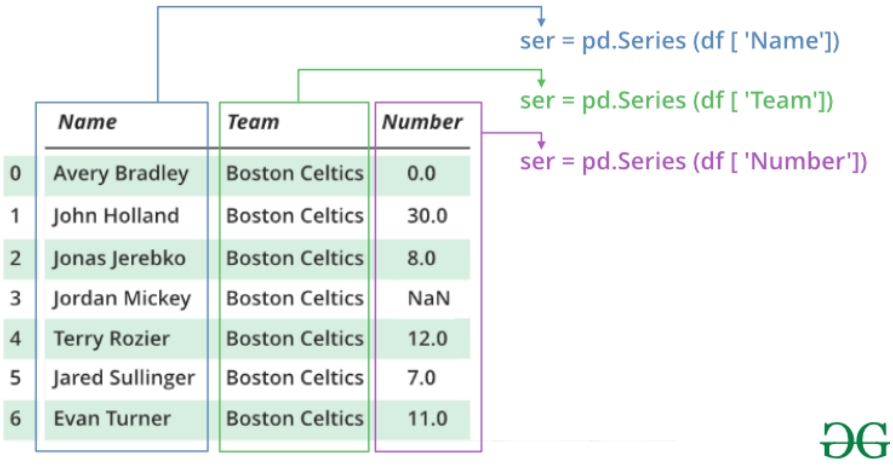
PROBLEM STATEMENT : Program on Pandas in Python

THEORY :

Pandas is a popular open-source library in Python for data manipulation and analysis. It provides data structures and functions for efficiently handling structured data, such as tables or spreadsheets, in a convenient and flexible manner.

Some of the key features of Pandas include:

1. **Data structures** : Pandas has two main data structures - Series and DataFrame. A Series is a one-dimensional array-like object that can hold any data type, while a DataFrame is a two-dimensional table-like data structure that consists of rows and columns, similar to a spreadsheet.
    A. **Series** :
    Pandas Series is a one-dimensional labeled array capable of holding data of any type (integer, string, float, python objects, etc.). The axis labels are collectively called indexes. Pandas Series is nothing but a column in an excel sheet. Labels need not be unique but must be a hashable type. The object supports both integer and label-based indexing and provides a host of methods for performing operations involving the index.



A Pandas Series will be created by loading the datasets from existing storage, storage can be SQL Database, CSV file, an Excel file. Pandas Series can be created from the lists, dictionary, and from a scalar value etc.

Example:

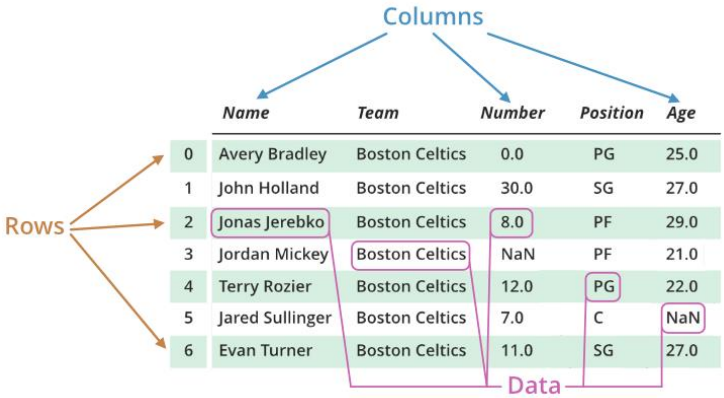
```
import pandas as pd
import numpy as np
ser = pd.Series()
print(ser)
data = np.array(['g', 'e', 'e', 'k', 's'])
ser = pd.Series(data)
print(ser)
```

OUTPUT :

```
Series([], dtype: float64)
0    g
1    e
2    e
3    k
4    s
dtype: object
```

B. **DATAFRAME** :

Pandas DataFrame is a two-dimensional size-mutable, potentially heterogeneous tabular data structure with labeled axes (rows and columns). A Data frame is a two-dimensional data structure, i.e., data is aligned in a tabular fashion in rows and columns. Pandas DataFrame consists of three principal components, the data, rows, and columns.



	Name	Team	Number	Position	Age
0	Avery Bradley	Boston Celtics	0.0	PG	25.0
1	John Holland	Boston Celtics	30.0	SG	27.0
2	Jonas Jerebko	Boston Celtics	8.0	PF	29.0
3	Jordan Mickey	Boston Celtics	NaN	PF	21.0
4	Terry Rozier	Boston Celtics	12.0	PG	22.0
5	Jared Sullinger	Boston Celtics	7.0	C	NaN
6	Evan Turner	Boston Celtics	11.0	SG	27.0

A Pandas DataFrame will be created by loading the datasets from existing storage, storage can be SQL Database, CSV file, an Excel file. Pandas DataFrame can be created from the lists, dictionary, and from a list of dictionaries, etc.

EXAMPLE :

```
import pandas as pd
df = pd.DataFrame()
print(df)
lst = ['Geeks', 'For', 'Geeks', 'is', 'portal', 'for', 'Geeks']
df = pd.DataFrame(lst)
print(d
```

OUTPUT :

```
Empty DataFrame
Columns: []
Index: []
        0
0   Geeks
1     For
2   Geeks
3      is
4  portal
5     for
6   Geeks
```

2. **Data manipulation** : Pandas provides various functions for manipulating and transforming data, such as filtering, sorting, grouping, merging, reshaping, and aggregating. These operations can be performed on entire datasets or subsets of the data.
3. **Missing data handling** : Pandas has built-in functions for handling missing data, such as filling missing values with a default value or interpolating missing values based on neighboring values.
4. **Time-series data analysis** : Pandas provides support for time-series data analysis, including date and time manipulation, time-zone handling, and frequency conversion.
5. **Input/output** : Pandas supports reading and writing data in various formats, such as CSV, Excel, SQL databases, and JSON.
6. **Visualization** : Pandas has built-in support for data visualization, allowing users to create various types of plots and charts using the Matplotlib library.

To use Pandas in Python, you can install it using the pip package manager by running pip install pandas in your terminal or command prompt. Once installed, you can import it in your Python code using import pandas as pd.

Advantages of Pandas in Python :

- Fast and efficient for manipulating and analyzing data.
- Data from different file objects can be loaded.
- Easy handling of missing data (represented as NaN) in floating point as well as non-floating point data
- Size mutability: columns can be inserted and deleted from DataFrame and higher dimensional objects
- Data set merging and joining.
- Flexible reshaping and pivoting of data sets
- Provides time-series functionality.
- Powerful group by functionality for performing split-apply-combine operations on data sets

CODE:

```
 import pandas as pd
import numpy as np
series1 = pd.Series([1, 2, 3, 4, 5])
series2 = pd.Series([10, 20, 30, 40, 50])
add_result = series1 + series2
print("Addition Result:\n", add_result)

sub_result = series1 - series2
print("Subtraction Result:\n", sub_result)

mult_result = series1 * series2
print("Multiplication Result:\n", mult_result)

div_result = series1 / series2
print("Division Result:\n", div_result)

series1 = pd.Series([1, 2, 3, 4, 5])
series2 = pd.Series([1, 3, 2, 5, 4])
comp_result = series1 == series2
print("Comparison Result:\n", comp_result)

my_dict = {'a': 100, 'b': 200, 'c': 300, 'd': 400, 'e': 500}
dict_series = pd.Series(my_dict)
print("Dictionary to Series Conversion:\n", dict_series)

my_array = np.array([10, 20, 30, 40, 50])
array_series = pd.Series(my_array)
print("NumPy Array to Series Conversion:\n", array_series)
```

Output:

Addition Result:

0 11

1 22

2 33

3 44

4 55

dtype: int64

Subtraction Result:

0 -9

1 -18

2 -27

3 -36

4 -45

dtype: int64

Multiplication Result:

0 10

1 40

2 90

3 160

4 250

dtype: int64

Division Result:

0 0.1

1 0.1

2 0.1

3 0.1

4 0.1

dtype: float64

Comparison Result:

0 True

1 False

2 False

3 True

4 False

dtype: bool

Dictionary to Series Conversion:

a 100

b 200

c 300

d 400

e 500

dtype: int64

NumPy Array to Series Conversion:

0 10

1 20

2 30

3 40

4 50

dtype: int64

CODE:

```
import pandas as pd
df = pd.read_csv("filename.csv")
print(“before csv:\n”)
print(df)
df.fillna(-999, inplace=True)
df.dropna(how='all', inplace=True)
print(“After csv:\n”)
print(df)
```

Output:

Before csv:

A,B,C,D

1,2,,4

5,,,7

8,9,10,

After csv:

```
  A  B     C  D
0  1  2 -999.0  4
1  5 -999.0  -999  7
2  8  9  10.0 -999
```

CODE:

```
import pandas as pd
df1 = pd.DataFrame({
   'id': [1, 2, 3, 4, 5],
   'name': ['Shruti', 'Niyati', 'Ujjwal', 'Kaushik', 'Jagjeet'],
   'age': [22, 33, 25, 27, 29]
})
df2 = pd.DataFrame({
   'id': [1, 2, 4, 5, 6],
   'salary': [50000, 60000, 70000, 80000, 90000],
   'department': ['Sales', 'Marketing', 'IT', 'Finance', 'HR']
})
merged_df = pd.merge(df1, df2, on='id')
print('Merged DataFrame on the basis of id:')
print(merged_df)
merged_df_outer = pd.merge(df1, df2, on='id', how='outer')
print('Merged DataFrame using outer method:')
print(merged_df_outer)
merged_df_inner = pd.merge(df1, df2, on='id', how='inner')
print('Merged DataFrame using inner method:')
print(merged_df_inner)
```

Output:

Merged DataFrame on the basis of id:

	id	name	age	salary	department
0	1	Shruti	22	50000	Sales
1	2	Niyati	33	60000	Marketing
2	4	Kaushik	27	70000	Finance
3	5	Jagjeet	29	80000	HR

Merged DataFrame using outer method:

	id	name	age	salary	department
0	1	Shruti	22.0	50000.0	Sales
1	2	Niyati	33.0	60000.0	Marketing
2	3	Ujjwal	25.0	NaN	NaN
3	4	Kaushik	27.0	70000.0	Finance
4	5	Jagjeet	29.0	80000.0	HR
5	6	NaN	NaN	90000.0	IT

Merged DataFrame using inner method:

	id	name	age	salary	department
0	1	Shruti	22	50000	Sales
1	2	Niyati	33	60000	Marketing
2	4	Kaushik	27	70000	Finance
3	5	Jagjeet	29	80000	HR