EXPERIMENT - 3(A)

PANDAS LIBRARY - DATA PREPROCESSING

Aim:

To understand the importance of data preprocessing in data science

Procedure:

- Upload the given dataset(csv file) and read it
- Import necessities such as numpy ,pandas
- Now, process the data, find missing values and replace it with mean(),mode(),median()

Program:

```
from google.colab import files
uploaded=files.upload()
import matplotlib.pyplot as plt
import pandsa as pd
import numpy as np
file=next(iter(uploaded))
df=pd.read_csv(file)
df.Country.fillna(df.Country.mode()[@],inplace=True)
df.Age.fillna(df.Age.median(),inplace=True)
df.Salary.fillna(round(df.Salary.mean()),inplace=True)
pd.get_dummies(df.Country)
df.Purchased.replace(['Yes','No'],[1,0],inplace=True)
df

Country Age Salary Purchased

0 France 44.0 7200.0 0

1 Spain 27.0 48000.0 1

2 Germany 30.0 54000.0 0

3 Spain 38.0 61000.0 0

4 Germany 40.0 63778.0 1

5 France 35.0 58000.0 1

6 Spain 38.0 52000.0 0

7 France 48.0 79000.0 1

8 Germany 50.0 83000.0 0

9 France 37.0 67000.0 1
```

Result:

Thus the python program to find missed values and replacing it was executed and verified

EXPERIMENT - 3(B)

PANDAS LIBRARY - HANDLING MISSING VALUES

Aim:

To handle and analyze missing and inappropriate data in dataset

Procedure:

• Upload the csv file and read it

- Import necessities such as pandas, numpy
- Now check for missing and inappropriate data to replace with appropriate data

Program:



Result:

Thus we find missing and inappropriate data and replaced with appropriate ones

EXPERIMENT-3(C)

PANDAS LIBRARY - CREATE CSV FILE

Aim:

To create a dataset and make it as csv file and handle inappropriate values

Procedure:

- Create a dataset
- Import pandas and numpy and with its help, create dataframe and change it to csv
- Read the dataset and handle inappropriate and missing values

Program:

Place ID Name Country Type Average Rating Entry Fee Best Season 4.7 Eiffel Tower France Monument 25.0 Spring Grand Canyon 2 USA Natural Wonder 4.8 35.0 Fall 2 3 Great Barrier Reef Australia Natural Wonder 4.9 NaN NaN 4.6 4 5 Mount Fuji Mountain 0.0 Japan Autumn 4.8 5 Island 10.0 6 Santorini Greece Summer Waterfall 4.7 NaN 6 Niagara Falls Canada All year 8 Petra Jordan Historical 46 70.0 Winter 9 Banff National Park Canada National Park 4.9 20.0 Colosseum Spring

```
os dff.info()
    <class 'pandas.core.frame.DataFrame'>
    RangeIndex: 10 entries, 0 to 9
    Data columns (total 7 columns):
     # Column
                        Non-Null Count Dtype
                                        int64
     0 Place_ID
                        10 non-null
                        10 non-null
                                        object
         Name
         Country
                        10 non-null
                                        object
         Туре
                        10 non-null
                                        object
         Average_Rating 8 non-null
                                        float64
                        8 non-null
                                        float64
     5 Entry_Fee
        Best_Season
                        10 non-null
                                        obiect
    dtypes: float64(2), int64(1), object(4)
    memory usage: 692.0+ bytes
```

```
dff.Average_Rating.fillna(df.Average_Rating.median(),inplace=True)
dff.Entry_Fee.fillna(df.Entry_Fee.median(),inplace=True)
dff
```

	Place_ID	Name	Country	Туре	Average_Rating	Entry_Fee	Best_Season
0	1	Eiffel Tower	France	Monument	4.70	25.0	Spring
1	2	Grand Canyon	USA	Natural Wonder	4.80	35.0	Fall
2	3	Great Barrier Reef	Australia	Natural Wonder	4.90	22.5	Summer
3	4	Machu Picchu	Peru	Historical	4.75	45.0	Spring
4	5	Mount Fuji	Japan	Mountain	4.60	0.0	Autumn
5	6	Santorini	Greece	Island	4.80	10.0	Summer
6	7	Niagara Falls	Canada	Waterfall	4.70	22.5	All year
7	8	Petra	Jordan	Historical	4.60	70.0	Winter
8	9	Banff National Park	Canada	National Park	4.90	20.0	Summer
9	10	Colosseum	Italy	Monument	4.75	18.0	Spring

```
[] dff.info()
     <class 'pandas.core.frame.DataFrame'>
RangeIndex: 10 entries, 0 to 9
     Data columns (total 7 columns):
     # Column
                         Non-Null Count Dtype
     0 Place_ID
                            10 non-null
                                                int64
                             10 non-null
                                                object
          Name
          Country
                             10 non-null
                                                object
          Type
                            10 non-null
                                                object
      4 Average_Rating 10 non-null
5 Entry_Fee 10 non-null
6 Best_Season 10 non-null
                                                float64
float64
                                                object
     dtypes: float64(2), int64(1), object(4)
     memory usage: 692.0+ bytes
```

Result:

Thus the python program to create own dataset and change it to csv file is executed and verified