Ex.No.6 **Data Wrangling**

Aim:

To do Data Wrangling functions

Description:

Data wrangling is the task in data science and analysis which includes operations like: Data Sorting: To rearrange values in ascending or descending order.
Data Filtration: To create a subset of available data.
Data Reduction: To eliminate or replace unwanted values.
Data Access: To read or write data files.
Data Processing: To perform aggregation, statistical, and similar operations on specific values.

1. Using join function to join two DataFrames.
2. Using combine function to combine two DataFrames.
3. Using merge function to merge two DataFrames.
4. Using replace function to replace the NaN values by average value.
5. Filtering and dropping the rows and rows and columns respectively.
6. Using concat function to concatenate two DataFrames.
7. Using melt function to reshape the DataFrame dimention.
8. Using groupby function to group the data set.
9. Using duplicated function to remove duplicated rows in the DataFram
10. Using merge function to merge two DataFrame data sets.

PROGRAM:

```
import pandas as pd

data1 = {'Name': ['Jai', 'Princi', 'Gaurav',
'Anuj', 'Ravi', 'Natasha', 'Tom', 'Rovana', 'Riya'],
'Roll No': [4,8,2,1,9,7,14,11,10],
'Age': [17, 17, 18, 17, 18, 17,19,16, 17],
'Gender': ['M', 'F', 'M', 'M', 'M', 'F','F','M', 'F']}

data2 = {'Name': ['Kelly', 'Natasha', 'Jack', 'Stacy',
'Stark', 'Loki', 'Rovana', 'Tom'],
'Roll No': [5,7,3,12,13,6,11,14],
'Age': [19, 17,16, 20, 17, 18, 16, 19],
'Gender': ['F','F' ,'M', 'F', 'M', 'M', 'F', 'M'],
'Marks': [95,71, 76, 94, 'NaN', 80,83, 68]}
```

```
marks = {'Marks': [80, 76, 'NaN', 74, 66,71,68,83, 'NaN']}

df1=  pd.DataFrame(data1)
df2=  pd.DataFrame(data2)
marks = pd.DataFrame(marks)

print("\nOriginal DataFrame 1:\n",df1)
print("\nOriginal DataFrame 2:\n",df1)
print("\nMarks:\n",marks)
```

```
df1 = df1.join(marks) print("\nDataFrame
1:\n",df1)
```

```
# Compute average c
= avg = 0
for ele in df1['Marks']: if
    str(ele).isnumeric():
        c += 1
        avg += ele
avg/= c

# Replace missing values
df1 = df1.replace(to_replace="NaN",value=avg) df2 =
df2.replace(to_replace="NaN",value=avg)   #   Display
data
print("\nReplacing NaN with Average marks:\nData Frame 1\n",df1)
print("\n\nData Frame 2\n",df2)
```

```
def myfunc(a, b):
    return a if a > b else b
df_combined  =  df1['Marks'].combine(df2['Marks'],  myfunc)
```

```
# Print the result
print("\nCombining the above two DataFrames using combine function with some condition:\n", df_combined)

---

newdf = df1.merge(df2, how='right')
print("\nMerge operation:\n",newdf)

---

df3 = pd.concat([df1,df2])
print("\nConcatenated DataFrame using cancat function:\n",df3)

---

# Group the data
grouped = df3.groupby('Age')
print("\nGroup by age 17:\n",grouped.get_group(17))

---

print("\nOriginal DataFrame:\n",df3)
#reshape DataFrame from wide format to long format
df = pd.melt(df3, id_vars='Roll No', value_vars=['Gender', 'Marks']) #view
updated DataFrame
print("\nReshaped Data Frame:\n",df)

---

# Filter top scoring students
df3=df3[df3['Marks'] >= 75] print("\nAfter
Filtering function:\n",df3) # Remove age
row
df3 = df3.drop(['Age'],axis=1)

# Display data
print("\nAfter Dropping function:\n",df3)

---
```

```
print("\nOriginal DataFrame:\n",df3)
# Here df.duplicated() list duplicate Entries in Rollno.
# So that ~(NOT) is placed in order to get non duplicate values. non_duplicate
=df3[~df3.duplicated('Roll No')]
#printing non-duplicate values
print("\nRemoved duplicated rows:\n",non_duplicate)
```

OUTPUT:

Original DataFrame 1:

	Name	Roll No	Age	Gender
0	Jai	4	17	M
1	Princi	8	17	F
2	Gaurav	2	18	M
3	Anuj	1	17	M
4	Ravi	9	18	M
5	Natasha	7	17	F
6	Tom	14	19	F
7	Rovana	11	16	M
8	Riya	10	17	F

Original DataFrame 2:

	Name	Roll No	Age	Gender	Marks
0	Kelly	5	19	F	95
1	Natasha	7	17	F	71
2	Jack	3	16	M	76
3	Stacy	12	20	F	94
4	Stark	13	17	M	NaN
5	Loki	6	18	M	80
6	Rovana	11	16	F	83
7	Tom	14	19	M	68

Marks:

	Marks
0	80
1	76
2	NaN
3	74
4	66
5	71
6	68
7	83
8	NaN

DataFrame 1:

	Name	Roll No	Age	Gender	Marks
0	Jai	4	17	M	80
1	Princi	8	17	F	76
2	Gaurav	2	18	M	NaN
3	Anuj	1	17	M	74
4	Ravi	9	18	M	66
5	Natasha	7	17	F	71
6	Tom	14	19	F	68
7	Rovana	11	16	M	83
8	Riya	10	17	F	NaN

Replacing NaN with Average marks:

Data Frame 1

	Name	Roll No	Age	Gender	Marks
0	Jai	4	17	M	80.0
1	Princi	8	17	F	76.0
2	Gaurav	2	18	M	74.0
3	Anuj	1	17	M	74.0
4	Ravi	9	18	M	66.0
5	Natasha	7	17	F	71.0
6	Tom	14	19	F	68.0
7	Rovana	11	16	M	83.0
8	Riya	10	17	F	74.0

Data Frame 2

	Name	Roll No	Age	Gender	Marks
0	Kelly	5	19	F	95.0
1	Natasha	7	17	F	71.0
2	Jack	3	16	M	76.0
3	Stacy	12	20	F	94.0
4	Stark	13	17	M	74.0
5	Loki	6	18	M	80.0
6	Rovana	11	16	F	83.0
7	Tom	14	19	M	68.0

Combining the above two DataFrames using combine function with some condition:

```
0    95.0
1    76.0
2    76.0
3    94.0
4    74.0
5    80.0
6    83.0
7    83.0
8     NaN
Name: Marks, dtype: float64
```

Merge operation:

	Name	Roll No	Age	Gender	Marks
0	Kelly	5	19	F	95.0
1	Natasha	7	17	F	71.0
2	Jack	3	16	M	76.0
3	Stacy	12	20	F	94.0
4	Stark	13	17	M	74.0
5	Loki	6	18	M	80.0
6	Rovana	11	16	F	83.0
7	Tom	14	19	M	68.0

Concatenated DataFrame using cancat function: Name Roll No Age Gender Marks

0	Jai	4	17	M	80.0
1	Princi	8	17	F	76.0
2	Gaurav	2	18	M	74.0
3	Anuj	1	17	M	74.0
4	Ravi	9	18	M	66.0
5	Natasha	7	17	F	71.0
6	Tom	14	19	F	68.0
7	Rovana	11	16	M	83.0
8	Riya	10	17	F	74.0
0	Kelly	5	19	F	95.0
1	Natasha	7	17	F	71.0
2	Jack	3	16	M	76.0
3	Stacy	12	20	F	94.0
4	Stark	13	17	M	74.0
5	Loki	6	18	M	80.0
6	Rovana	11	16	F	83.0
7	Tom	14	19	M	68.0

Group by age 17:

	Name	Roll No	Age	Gender	Marks
0	Jai	4	17	M	80.0
1	Princi	8	17	F	76.0
3	Anuj	1	17	M	74.0
5	Natasha	7	17	F	71.0
8	Riya	10	17	F	74.0
1	Natasha	7	17	F	71.0
4	Stark	13	17	M	74.0

Original DataFrame:

	Name	Roll No	Age	Gender	Marks
0	Jai	4	17	M	80.0
1	Princi	8	17	F	76.0
2	Gaurav	2	18	M	74.0
3	Anuj	1	17	M	74.0
4	Ravi	9	18	M	66.0
5	Natasha	7	17	F	71.0
6	Tom	14	19	F	68.0
7	Rovana	11	16	M	83.0
8	Riya	10	17	F	74.0
0	Kelly	5	19	F	95.0
1	Natasha	7	17	F	71.0
2	Jack	3	16	M	76.0
3	Stacy	12	20	F	94.0
4	Stark	13	17	M	74.0
5	Loki	6	18	M	80.0
6	Rovana	11	16	F	83.0
7	Tom	14	19	M	68.0

Reshaped Data Frame:

	Roll No	variable	value
0	4	Gender	M
1	8	Gender	F
2	2	Gender	M
3	1	Gender	M
4	9	Gender	M
5	7	Gender	F
6	14	Gender	F
7	11	Gender	M
8	10	Gender	F
9	5	Gender	F
10	7	Gender	F
11	3	Gender	M
12	12	Gender	F
13	13	Gender	M
14	6	Gender	M
15	11	Gender	F
16	14	Gender	M
17	4	Marks	80.0
18	8	Marks	76.0
19	2	Marks	74.0
20	1	Marks	74.0
21	9	Marks	66.0
22	7	Marks	71.0
23	14	Marks	68.0
24	11	Marks	83.0
25	10	Marks	74.0
26	5	Marks	95.0
27	7	Marks	71.0
28	3	Marks	76.0
29	12	Marks	94.0
30	13	Marks	74.0
31	6	Marks	80.0
32	11	Marks	83.0
33	14	Marks	68.0

After Filtering function:

	Name	Roll No	Age	Gender	Marks
0	Jai	4	17	M	80.0
1	Princi	8	17	F	76.0
7	Rovana	11	16	M	83.0
0	Kelly	5	19	F	95.0
2	Jack	3	16	M	76.0
3	Stacy	12	20	F	94.0
5	Loki	6	18	M	80.0
6	Rovana	11	16	F	83.0

After Dropping function:

	Name	Roll No	Gender	Marks
0	Jai	4	M	80.0
1	Princi	8	F	76.0
7	Rovana	11	M	83.0
0	Kelly	5	F	95.0
2	Jack	3	M	76.0
3	Stacy	12	F	94.0
5	Loki	6	M	80.0
6	Rovana	11	F	83.0

Original DataFrame:

	Name	Roll No	Gender	Marks
0	Jai	4	M	80.0
1	Princi	8	F	76.0
7	Rovana	11	M	83.0
0	Kelly	5	F	95.0
2	Jack	3	M	76.0
3	Stacy	12	F	94.0
5	Loki	6	M	80.0
6	Rovana	11	F	83.0

Removed duplicated rows:

	Name	Roll No	Gender	Marks
0	Jai	4	M	80.0
1	Princi	8	F	76.0
7	Rovana	11	M	83.0
0	Kelly	5	F	95.0
2	Jack	3	M	76.0
3	Stacy	12	F	94.0
5	Loki	6	M	80.0

Result:

The programs were run successfully