#### 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
```



```
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)
                                          Edit with WPS Office
```

```
# 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
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)

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### OUTPUT:

### Original DataFrame 1:

	Nam	Roll No	Age	Gende	
0	е		4	r 17	M
	Jai				
1	Princi		8	17	F
2	Gaurav		2	18	M
2 3 4 5	Anuj		1	17	M
4	Ravi		9	18	M
5	Natasha		7	17	F
6	Tom		1	19	F
			4		
7	Rovana		1	16	M
			1		
8	Riya		1	17	F
			0		

#### Original DataFrame 2:

	Name	Roll No Ag	je Gen	der 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



Da <sup>-</sup>	taFrame 1:					
	Nam	Roll No	Ag	e Gende	Marks	
0	е		4	r 17	M	80
	Jai					
1	Princi		8	17	F	76
2	Gaurav		2	18	M	NaN
3 4	Anuj		1	17	M	74
4	Ravi		9	18	M	66
5	Natasha		7	17	F	71
6	Tom		1	19	F	68
			4			
7	Rovana		1	16	M	83
			1			
8	Riya		1	17	F	NaN
			0			

## Replacing NaN with Average marks:

Dat	-	-r	$^{\circ}$	n	⊃ ¹I
υa	La		aı.	115	_

	Nam	Roll No	Ag	e Gende	Marks	
0	е		4	r 17	M	80.0
	Jai					
1	Princi		8	17	F	76.0
2	Gaurav		2	18	M	74.0
2	Anuj		1	17	M	74.0
4	Ravi		9	18	M	66.0
5	Natasha		7	17	F	71.0
6	Tom		1	19	F	68.0
			4			
7	Rovana		1	16	M	83.0
			1			
8	Riya		1	17	F	74.0
			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:

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



Merg	operatio					
е	n:	Roll No	Ag	e Gender	Marks	
	Name					
0	Kelly		5	19	F	95.0
1	Natash		7	17	F	71.0
	а					
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

Gro	up by age	17:				
		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	Natash		7	17	F	71.0
	a					
8	Riya		10	17	F	74.0
1	Natash		7	17	F	71.0
	a					
4	Stark		13	17	M	74.0



Original DataFrame:							
	Name	Roll No Ag	e 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:

1100111	apoubutuii	41110.	
0	Roll No vari	able value	
0	4	Gender	М
1	4 8 2 1	Gender	
2	2	Gender	M
3	1	Gender	M
4	9	Gender	M
5	/	Gender	F
6	14	Gender	F
1 2 3 4 5 6 7 8 9	11 10 5 7	Gender	M
8	10	Gender	F
9	5	Gender	F
10	7	Gender	F
11	3	Gender	M
12	3 12 13	Gender Gender	F
13	13	Gender	M
10 11 12 13 14 15	6	Gender Gender	F M M F F M F M M
15	11	Gender	F
16	14	Gender	M
17	14 4 8 2 1	Marks	80.0
18	8	Marks	76.0
19	2	Marks	74.0
20	1	Marks	76.0 74.0 74.0
21	9 7	Marks	66.0
22		Marks	71.0
17 18 19 20 21 22 23 24 25 26 27 28 29 30	14	Marks Marks Marks	68.0
24	11 10 5 7	Marks	83 0
25	10	Marks	74.0
26	5	Marks	95.0
27	7	Marks	71.0
28	3 12 13	Marks	74.0 95.0 71.0 76.0 94.0
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 Mark

	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:

	Nam	Roll No Gender	Mark	
0	e Jai	4	s 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

