

Ex.No.6**DataWrangling****Aim:**

TodoDataWranglingfunctions

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 columns respectively.
6. Using concat function to concatenate two DataFrames.
7. Using melt function to reshape the DataFrame dimension.
8. Using groupby function to group the dataset.
9. Using duplicated function to remove duplicated rows in the DataFrame.
10. Using merge function to merge two DataFrame datasets.

PROGRAM:

```
import pandas as pd

data1 = {'Name': ['Jai', 'Princi', 'Gaurav',
                 'Anuj', 'Ravi', 'Natasha', 'Tom', 'Rovana', 'Riya'],
         'RollNo': [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'],
         'RollNo': [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("\nOriginalDataFrame1:\n",df1)
```

```
print("\nOriginalDataFrame2:\n",df2)
```

```
print("\nMarks:\n",marks)
```

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

```
#Computeaveragec
```

```
=avg=0
```

```
for ele in df1['Marks']:if
```

```
    str(ele).isnumeric():
```

```
        c+=1
```

```
        avg+=ele
```

```
avg/= c
```

```
#Replacemissingvalues
```

```
df1 = df1.replace(to_replace="NaN",value=avg) df2 =
```

```
df2.replace(to_replace="NaN",value=avg) # Display
```

```
data
```

```
print("\nReplacingNaNwithAveragemarks:\nDataFrame1\n",df1)
```

```
print("\n\nData Frame 2\n",df2)
```

```
defmyfunc(a,b):
```

```
    returnaifa>belseb
```

```
df_combined=df1['Marks'].combine(df2['Marks'],myfunc)
```

```
#Printtheresult
print("\nCombiningtheabovetwoDataFramesusingcombinefunctionwithsomecondition:\n",df_combined)
```

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

```
df3=pd.concat([df1,df2])
print("\nConcatenatedDataFrameusingconcatfunction:\n",df3)
```

```
#Groupthedata
grouped= df3.groupby('Age')
print("\nGroupbyage17:\n",grouped.get_group(17))
```

```
print("\nOriginalDataFrame:\n",df3)
#reshapeDataFramefromwideformattolongformat
df=pd.melt(df3,id_vars='RollNo', value_vars=['Gender', 'Marks'])#view
updatedDataFrame
print("\nReshapedData Frame:\n",df)
```

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

```
#Displaydata
print("\nAfterDroppingfunction:\n",df3)
```

```

print("\nOriginalDataFrame:\n",df3)
#Heredf.duplicated()listduplicateEntriesinRollno.
#Sothat~(NOT)isplacedinordertogetnonduplicatevalues.non_duplicate
=df3[~df3.duplicated('RollNo')]
#printing non-duplicate values
print("\nRemovedduplicatedrows:\n",non_duplicate)

```

OUTPUT:

OriginalDataFrame1:

	Name	RollNo	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

OriginalDataFrame2:

	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

DataFrame1:

	Name	RollNo	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

ReplacingNaNwithAveragemarks:

DataFrame1

RollNo	Name	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

DataFrame2

	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

CombiningtheabovetwoDataFramesusingcombinefunctionwithsome 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	RollNo	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

ConcatenatedDataFrameusingcancatfunction:NameRollNoAge
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:	No	Age	Gender	Marks
	Name	Roll				
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

OriginalDataFrame:

	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

ReshapedDataFrame:

	RollNo	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

AfterFilteringfunction:

	Name	RollNo	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

AfterDroppingfunction:

	Name	RollNo	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

OriginalDataFrame:

	Name	RollNo	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

Removedduplicatedrows:

	Name	RollNo	Gender	Marks
0	Jai		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:

Theprogramswererunsuccessfully