### In [10]:

```
import pandas as pd
import numpy as np

df=pd.read_csv("C:\\Users\\Nancy\\Desktop\\Nancy_Assign2.csv")

df.head()
df
```

### Out[10]:

	Gender	Math_score	Reading_score	Writing_score	Placement_score	Placement_offer_coun
0	Female	15.0	90.0	77.0	88.0	3
1	Female	5.0	82.0	78.0	93.0	(
2	Female	170.0	100.0	74.0	32.0	2
3	Male	200.0	89.0	32.0	NaN	
4	NaN	1.0	79.0	67.0	100.0	:
5	Female	56.0	79.0	33.0	80.0	2
6	Male	16.0	94.0	NaN	80.0	2
7	Female	NaN	90.0	68.0	83.0	3
8	Male	36.0	50.0	72.0	90.0	2
9	Male	69.0	90.0	79.0	78.0	,
10	Female	0.0	83.0	74.0	89.0	2
11	Female	6.0	87.0	79.0	98.0	:
12	Male	4.0	87.0	75.0	93.0	2
13	Male	150.0	43.0	65.0	21.0	,
14	Male	76.0	75.0	62.0	78.0	,
15	Female	3.0	46.0	74.0	93.0	2
16	Male	103.0	NaN	62.0	94.0	,
17	Male	26.0	78.0	100.0	87.0	2
18	Male	78.0	95.0	75.0	11.0	2
19	Female	82.0	82.0	NaN	76.0	(
20	Female	17.0	82.0	69.0	85.0	2
4						<b>•</b>

### In [19]:

```
from sklearn.preprocessing import LabelEncoder
le = LabelEncoder()
df['Gender'] = le.fit_transform(df['Gender'])
newdf=df
df
```

# Out[19]:

	Gender	Math score	Reading score	Writing score	Placement score	Placement offer count	Region
0	0	15.0	90.0	77.0	88.0	2	Nashik
1	0	5.0	82.0	78.0	93.0	2	Pune
2	0	93.0	93.0	74.0	85.0	2	Nashik
3	1	34.0	89.0	67.0	NaN	1	Nashik
4	2	1.0	79.0	67.0	100.0	3	Nashik
5	0	56.0	79.0	70.0	80.0	2	Nashik
6	1	16.0	94.0	NaN	80.0	2	Pune
7	0	NaN	90.0	68.0	83.0	2	Pune
8	1	36.0	82.0	72.0	90.0	2	Nashik
9	1	69.0	90.0	79.0	78.0	1	Nashik
10	0	0.0	83.0	74.0	89.0	2	Pune
11	0	6.0	87.0	79.0	98.0	3	Nashik
12	1	4.0	87.0	75.0	93.0	2	Pune
13	1	58.0	83.0	65.0	78.0	1	Pune
14	1	76.0	75.0	62.0	78.0	1	NaN
15	0	3.0	84.0	74.0	93.0	2	Nashik
16	1	39.0	NaN	62.0	94.0	2	Nashik
17	1	26.0	78.0	70.0	87.0	2	Nashik
18	1	78.0	95.0	75.0	81.0	2	Nashik
19	0	82.0	82.0	NaN	76.0	1	Pune
20	0	17.0	82.0	69.0	85.0	2	Nashik

### In [20]:

```
missing_values = ["Na", "na"]
df = pd.read_csv("C:\\Users\\Nancy\\Desktop\\Nancy_Assign2.csv")
df
```

# Out[20]:

	Gender	Math score	Reading score	Writing score	Placement score	Placement offer count	Region
0	Female	15.0	90.0	77.0	88.0	2	Nashik
1	Female	5.0	82.0	78.0	93.0	2	Pune
2	Female	93.0	93.0	74.0	85.0	2	Nashik
3	Male	34.0	89.0	67.0	NaN	1	Nashik
4	NaN	1.0	79.0	67.0	100.0	3	Nashik
5	Female	56.0	79.0	70.0	80.0	2	Nashik
6	Male	16.0	94.0	NaN	80.0	2	Pune
7	Female	NaN	90.0	68.0	83.0	2	Pune
8	Male	36.0	82.0	72.0	90.0	2	Nashik
9	Male	69.0	90.0	79.0	78.0	1	Nashik
10	Female	0.0	83.0	74.0	89.0	2	Pune
11	Female	6.0	87.0	79.0	98.0	3	Nashik
12	Male	4.0	87.0	75.0	93.0	2	Pune
13	Male	58.0	83.0	65.0	78.0	1	Pune
14	Male	76.0	75.0	62.0	78.0	1	NaN
15	Female	3.0	84.0	74.0	93.0	2	Nashik
16	Male	39.0	NaN	62.0	94.0	2	Nashik
17	Male	26.0	78.0	70.0	87.0	2	Nashik
18	Male	78.0	95.0	75.0	81.0	2	Nashik
19	Female	82.0	82.0	NaN	76.0	1	Pune
20	Female	17.0	82.0	69.0	85.0	2	Nashik

# In [12]:

ndf = df
ndf.fillna(0)

### Out[12]:

	Gender	Math_score	Reading_score	Writing_score	Placement_score	Placement_offer_coun
0	Female	15.0	90.0	77.0	88.0	{
1	Female	5.0	82.0	78.0	93.0	•
2	Female	170.0	100.0	74.0	32.0	2
3	Male	200.0	89.0	32.0	0.0	,
4	0	1.0	79.0	67.0	100.0	:
5	Female	56.0	79.0	33.0	80.0	2
6	Male	16.0	94.0	0.0	80.0	2
7	Female	0.0	90.0	68.0	83.0	:
8	Male	36.0	50.0	72.0	90.0	2
9	Male	69.0	90.0	79.0	78.0	,
10	Female	0.0	83.0	74.0	89.0	2
11	Female	6.0	87.0	79.0	98.0	:
12	Male	4.0	87.0	75.0	93.0	2
13	Male	150.0	43.0	65.0	21.0	,
14	Male	76.0	75.0	62.0	78.0	,
15	Female	3.0	46.0	74.0	93.0	2
16	Male	103.0	0.0	62.0	94.0	,
17	Male	26.0	78.0	100.0	87.0	2
18	Male	78.0	95.0	75.0	11.0	2
19	Female	82.0	82.0	0.0	76.0	(
20	Female	17.0	82.0	69.0	85.0	2

4 | ·

# In [20]:

```
m_v = df['Math_score'].mean()
df['Math_score'].fillna(value=m_v, inplace=True)
df
```

### Out[20]:

	Gender	Math_score	Reading_score	Writing_score	Placement_score	Placement_offer_coun
0	Female	15.00	90.0	77.0	88.0	3
1	Female	5.00	82.0	78.0	93.0	ť
2	Female	170.00	100.0	74.0	32.0	2
3	Male	200.00	89.0	32.0	NaN	,
4	NaN	1.00	79.0	67.0	100.0	:
5	Female	56.00	79.0	33.0	80.0	2
6	Male	16.00	94.0	NaN	80.0	2
7	Female	55.65	90.0	68.0	83.0	:
8	Male	36.00	50.0	72.0	90.0	2
9	Male	69.00	90.0	79.0	78.0	,
10	Female	0.00	83.0	74.0	89.0	2
11	Female	6.00	87.0	79.0	98.0	:
12	Male	4.00	87.0	75.0	93.0	2
13	Male	150.00	43.0	65.0	21.0	,
14	Male	76.00	75.0	62.0	78.0	,
15	Female	3.00	46.0	74.0	93.0	2
16	Male	103.00	NaN	62.0	94.0	,
17	Male	26.00	78.0	100.0	87.0	2
18	Male	78.00	95.0	75.0	11.0	2
19	Female	82.00	82.0	NaN	76.0	(
20	Female	17.00	82.0	69.0	85.0	2

**◆** 

In [23]:

ndf.replace(to\_replace = np.nan, value = -99)

# Out[23]:

	Gender	Math score	Reading score	Writing score	Placement score	Placement offer count	Region
0	Female	15.0	90.0	77.0	88.0	2	Nashik
1	Female	5.0	82.0	78.0	93.0	2	Pune
2	Female	93.0	93.0	74.0	85.0	2	Nashik
3	Male	34.0	89.0	67.0	-99.0	1	Nashik
4	<b>-</b> 99	1.0	79.0	67.0	100.0	3	Nashik
5	Female	56.0	79.0	70.0	80.0	2	Nashik
6	Male	16.0	94.0	-99.0	80.0	2	Pune
7	Female	35.7	90.0	68.0	83.0	2	Pune
8	Male	36.0	82.0	72.0	90.0	2	Nashik
9	Male	69.0	90.0	79.0	78.0	1	Nashik
10	Female	0.0	83.0	74.0	89.0	2	Pune
11	Female	6.0	87.0	79.0	98.0	3	Nashik
12	Male	4.0	87.0	75.0	93.0	2	Pune
13	Male	58.0	83.0	65.0	78.0	1	Pune
14	Male	76.0	75.0	62.0	78.0	1	-99
15	Female	3.0	84.0	74.0	93.0	2	Nashik
16	Male	39.0	-99.0	62.0	94.0	2	Nashik
17	Male	26.0	78.0	70.0	87.0	2	Nashik
18	Male	78.0	95.0	75.0	81.0	2	Nashik
19	Female	82.0	82.0	-99.0	76.0	1	Pune
20	Female	17.0	82.0	69.0	85.0	2	Nashik

# In [24]:

# ndf.dropna()

# Out[24]:

	Gender	Math score	Reading score	Writing score	Placement score	Placement offer count	Region
0	Female	15.0	90.0	77.0	88.0	2	Nashik
1	Female	5.0	82.0	78.0	93.0	2	Pune
2	Female	93.0	93.0	74.0	85.0	2	Nashik
5	Female	56.0	79.0	70.0	80.0	2	Nashik
7	Female	35.7	90.0	68.0	83.0	2	Pune
8	Male	36.0	82.0	72.0	90.0	2	Nashik
9	Male	69.0	90.0	79.0	78.0	1	Nashik
10	Female	0.0	83.0	74.0	89.0	2	Pune
11	Female	6.0	87.0	79.0	98.0	3	Nashik
12	Male	4.0	87.0	75.0	93.0	2	Pune
13	Male	58.0	83.0	65.0	78.0	1	Pune
15	Female	3.0	84.0	74.0	93.0	2	Nashik
17	Male	26.0	78.0	70.0	87.0	2	Nashik
18	Male	78.0	95.0	75.0	81.0	2	Nashik
20	Female	17.0	82.0	69.0	85.0	2	Nashik

# In [25]:

ndf.dropna(how = 'all')

### Out[25]:

	Gender	Math score	Reading score	Writing score	Placement score	Placement offer count	Region
0	Female	15.0	90.0	77.0	88.0	2	Nashik
1	Female	5.0	82.0	78.0	93.0	2	Pune
2	Female	93.0	93.0	74.0	85.0	2	Nashik
3	Male	34.0	89.0	67.0	NaN	1	Nashik
4	NaN	1.0	79.0	67.0	100.0	3	Nashik
5	Female	56.0	79.0	70.0	80.0	2	Nashik
6	Male	16.0	94.0	NaN	80.0	2	Pune
7	Female	35.7	90.0	68.0	83.0	2	Pune
8	Male	36.0	82.0	72.0	90.0	2	Nashik
9	Male	69.0	90.0	79.0	78.0	1	Nashik
10	Female	0.0	83.0	74.0	89.0	2	Pune
11	Female	6.0	87.0	79.0	98.0	3	Nashik
12	Male	4.0	87.0	75.0	93.0	2	Pune
13	Male	58.0	83.0	65.0	78.0	1	Pune
14	Male	76.0	75.0	62.0	78.0	1	NaN
15	Female	3.0	84.0	74.0	93.0	2	Nashik
16	Male	39.0	NaN	62.0	94.0	2	Nashik
17	Male	26.0	78.0	70.0	87.0	2	Nashik
18	Male	78.0	95.0	75.0	81.0	2	Nashik
19	Female	82.0	82.0	NaN	76.0	1	Pune
20	Female	17.0	82.0	69.0	85.0	2	Nashik

# In [26]:

# ndf.dropna(axis = 1)

### Out[26]:

	Math score	Placement offer count
0	15.0	2
1	5.0	2
2	93.0	2
3	34.0	1
4	1.0	3
5	56.0	2
6	16.0	2
7	35.7	2
8	36.0	2
9	69.0	1
10	0.0	2
11	6.0	3
12	4.0	2
13	58.0	1
14	76.0	1
15	3.0	2
16	39.0	2
17	26.0	2
18	78.0	2
19	82.0	1
20	17.0	2

# In [27]:

```
new_data = ndf.dropna(axis = 0, how = 'any')
new_data
```

### Out[27]:

	Gender	Math score	Reading score	Writing score	Placement score	Placement offer count	Region
0	Female	15.0	90.0	77.0	88.0	2	Nashik
1	Female	5.0	82.0	78.0	93.0	2	Pune
2	Female	93.0	93.0	74.0	85.0	2	Nashik
5	Female	56.0	79.0	70.0	80.0	2	Nashik
7	Female	35.7	90.0	68.0	83.0	2	Pune
8	Male	36.0	82.0	72.0	90.0	2	Nashik
9	Male	69.0	90.0	79.0	78.0	1	Nashik
10	Female	0.0	83.0	74.0	89.0	2	Pune
11	Female	6.0	87.0	79.0	98.0	3	Nashik
12	Male	4.0	87.0	75.0	93.0	2	Pune
13	Male	58.0	83.0	65.0	78.0	1	Pune
15	Female	3.0	84.0	74.0	93.0	2	Nashik
17	Male	26.0	78.0	70.0	87.0	2	Nashik
18	Male	78.0	95.0	75.0	81.0	2	Nashik
20	Female	17.0	82.0	69.0	85.0	2	Nashik

```
In [30]:
```

```
import pandas as pd
import numpy as np

df1 = pd.read_csv("C:\\Users\\Nancy\\Desktop\\heights.csv")
df1
```

#### Out[30]:

	name	neight
0	mohan	5.9
1	maria	5.2
2	sakib	5.1
3	tao	5.5
4	virat	4.9
5	vaishnav	6.2
6	yash	5.6
7	vanshika	7.1
8	siddharth	14.2
9	taabish	6.5
10	john	6.1
11	joseph	1.2
12	celina	5.5

### In [31]:

```
df1['height'].quantile(0.95)
```

#### Out[31]:

9.93999999999999

#### In [32]:

```
max_thresold = df1['height'].quantile(0.95)
max_thresold
```

#### Out[32]:

9.93999999999989

#### In [33]:

```
df1[df1['height']>max_thresold]
```

#### Out[33]:

	name	height
8	siddharth	14.2

```
In [34]:
min_thresold = df1['height'].quantile(0.05)
min_thresold
Out[34]:
3.420000000000001
In [35]:
df1[df1['height']<min_thresold]</pre>
Out[35]:
     name height
    joseph
              1.2
In [36]:
df1[(df1['height']<max_thresold) & (df1['height']>min_thresold)]
Out[36]:
       name height
  0
      mohan
                5.9
  1
       maria
                5.2
  2
       sakib
                5.1
  3
                5.5
         tao
  4
        virat
                4.9
     vaishnav
                6.2
  5
        yash
                5.6
     vanshika
                7.1
      taabish
                6.5
```

```
In [37]:
```

john

celina

6.1

5.5

10

12

```
df2 = df1[(df1['height']<max_thresold) & (df1['height']>min_thresold)]
df2.shape
```

```
Out[37]:
```

(11, 2)

#### In [38]:

#### df2.describe()

#### Out[38]:

	height
count	11.000000
mean	5.781818
std	0.656990
min	4.900000
25%	5.350000
50%	5.600000
75%	6.150000
max	7.100000

### In [23]:

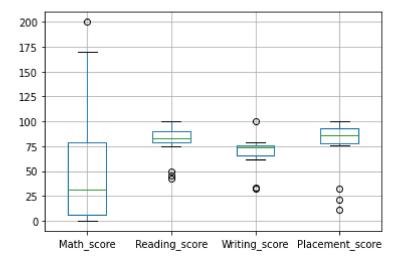
```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt

df=pd.read_csv("C:\\Users\\Nancy\\Desktop\\Nancy_Assign2.csv")

col = ['Math_score', 'Reading_score', 'Writing_score', 'Placement_score']
df.boxplot(col)
```

#### Out[23]:

#### <AxesSubplot:>



### In [27]:

```
(array([ 2,  3, 13, 16], dtype=int64),)
(array([ 8, 13, 15], dtype=int64),)
(array([3, 5], dtype=int64),)
```

#### In [43]:

```
import pandas as pd
df = pd.read_csv("C:\\Users\\Nancy\\Desktop\\heights.csv")
df
```

#### Out[43]:

	name	height
0	mohan	5.9
1	maria	5.2
2	sakib	5.1
3	tao	5.5
4	virat	4.9
5	vaishnav	6.2
6	yash	5.6
7	vanshika	7.1
8	siddharth	14.2
9	taabish	6.5
10	john	6.1
11	joseph	1.2
12	celina	5.5

#### In [44]:

```
df.describe()
```

### Out[44]:

	height
count	13.000000
mean	6.076923
std	2.814295
min	1.200000
25%	5.200000
50%	5.600000
75%	6.200000
max	14.200000

```
In [45]:
Q1 = df.height.quantile(0.25)
Q3 = df.height.quantile(0.75)
Q1, Q3
Out[45]:
(5.2, 6.2)
In [46]:
IQR = Q3 - Q1
IQR
Out[46]:
1.0
In [47]:
lower_limit = Q1 - 1.5*IQR
upper_limit = Q3 + 1.5*IQR
lower_limit, upper_limit
Out[47]:
(3.7, 7.7)
In [48]:
df[(df.height<lower_limit) | (df.height>upper_limit)]
Out[48]:
```

name height

8

11

siddharth

joseph

14.2

1.2

### In [49]:

```
df_no_outlier = df[(df.height>lower_limit) & (df.height<upper_limit)]
df_no_outlier</pre>
```

### Out[49]:

name	height
mohan	5.9
maria	5.2
sakib	5.1
tao	5.5
virat	4.9
vaishnav	6.2
yash	5.6
vanshika	7.1
taabish	6.5
john	6.1
celina	5.5
	mohan maria sakib tao virat vaishnav yash vanshika taabish john

### In [53]:

```
df.shape
```

### Out[53]:

(13, 2)

### In [52]:

```
df_no_outlier.shape
```

### Out[52]:

(11, 2)

### In [7]:

```
import pandas as pd
import numpy as np

df=pd.read_csv("C:\\Users\\Nancy\\Desktop\\Nancy_Assign2.csv")
df.head()
print(df)
```

	Gender	Math_score	Reading_score	Writing_score	Placement_score	\
0	Female	15.0	90.0	77.0	88.0	
1	Female	5.0	82.0	78.0	93.0	
2	Female	170.0	100.0	74.0	32.0	
3	Male	200.0	89.0	32.0	NaN	
4	NaN	1.0	79.0	67.0	100.0	
5	Female	56.0	79.0	33.0	80.0	
6	Male	16.0	94.0	NaN	80.0	
7	Female	NaN	90.0	68.0	83.0	
8	Male	36.0	50.0	72.0	90.0	
9	Male	69.0	90.0	79.0	78.0	
10	Female	0.0	83.0	74.0	89.0	
11	Female	6.0	87.0	79.0	98.0	
12	Male	4.0	87.0	75.0	93.0	
13	Male	150.0	43.0	65.0	21.0	
14	Male	76.0	75.0	62.0	78.0	
15	Female	3.0	46.0	74.0	93.0	
16	Male	103.0	NaN	62.0	94.0	
17	Male	26.0	78.0	100.0	87.0	
18	Male	78.0	95.0	75.0	11.0	
19	Female	82.0	82.0	NaN	76.0	
20	Female	17.0	82.0	69.0	85.0	

	Placement_offer_count	Region
0	8	Nashik
1	6	Pune
2	2	Nashik
3	1	Nashik
4	3	Nashik
5	2	Nashik
6	2	Pune
7	3	Pune
8	2	Nashik
9	1	Nashik
10	2	Pune
11	3	Nashik
12	2	Pune
13	1	Pune
14	1	NaN
15	2	Nashik
16	1	Nashik
17	2	Nashik
18	2	Nashik
19	0	Pune
20	2	Nashik

```
In [8]:
Q1 = df.Math_score.quantile(0.25)
Q3 = df.Math_score.quantile(0.75)
Q1, Q3
Out[8]:
(5.75, 79.0)
In [9]:
IQR = Q3 - Q1
IQR
Out[9]:
73.25
In [10]:
lower_limit = Q1 - 1.5*IQR
upper_limit = Q3 + 1.5*IQR
lower_limit, upper_limit
Out[10]:
(-104.125, 188.875)
In [12]:
df[(df.Math_score<lower_limit) | (df.Math_score>upper_limit)]
Out[12]:
   Gender
                      Reading_score Writing_score Placement_score Placement_offer_count
           Math_score
 3
      Male
                200.0
                               89.0
                                            32.0
                                                           NaN
                                                                                  1
```

### In [13]:

```
df_no_outlier = df[(df.Math_score>lower_limit) & (df.Math_score<upper_limit)]
df_no_outlier</pre>
```

### Out[13]:

	Gender	Math_score	Reading_score	Writing_score	Placement_score	Placement_offer_coun
0	Female	15.0	90.0	77.0	88.0	
1	Female	5.0	82.0	78.0	93.0	(
2	Female	170.0	100.0	74.0	32.0	2
4	NaN	1.0	79.0	67.0	100.0	:
5	Female	56.0	79.0	33.0	80.0	2
6	Male	16.0	94.0	NaN	80.0	2
8	Male	36.0	50.0	72.0	90.0	2
9	Male	69.0	90.0	79.0	78.0	,
10	Female	0.0	83.0	74.0	89.0	2
11	Female	6.0	87.0	79.0	98.0	:
12	Male	4.0	87.0	75.0	93.0	2
13	Male	150.0	43.0	65.0	21.0	,
14	Male	76.0	75.0	62.0	78.0	,
15	Female	3.0	46.0	74.0	93.0	2
16	Male	103.0	NaN	62.0	94.0	,
17	Male	26.0	78.0	100.0	87.0	2
18	Male	78.0	95.0	75.0	11.0	2
19	Female	82.0	82.0	NaN	76.0	(
20	Female	17.0	82.0	69.0	85.0	2

4

### In [14]:

df.shape

### Out[14]:

(21, 7)

### In [15]:

df\_no\_outlier.shape

### Out[15]:

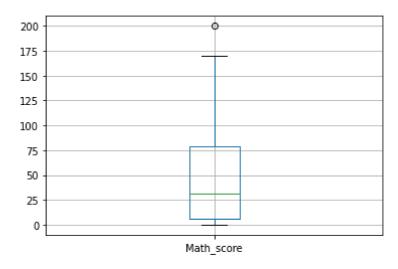
(19, 7)

# In [20]:

```
df.boxplot("Math_score")
```

# Out[20]:

# <AxesSubplot:>



#### In [35]:

```
import pandas as pd
import numpy as np

df=pd.read_csv("C:\\Users\\Nancy\\Desktop\\Nancy_Assign2.csv")

df.head()

Q1 = np.nanpercentile(df["Math_score"],25)
Q3 = np.nanpercentile(df["Math_score"],75)
print(Q1, Q3)

Gender Math_score Reading_score Writing_score Placement_score \
Outside Samelan = 15.00
```

	Gender	Math_score	Reading_score	Writing_score	Placement_score
0	Female	15.0	90.0	77.0	88.0
1	Female	5.0	82.0	78.0	93.0
2	Female	170.0	100.0	74.0	32.0
3	Male	200.0	89.0	32.0	NaN
4	NaN	1.0	79.0	67.0	100.0
5	Female	56.0	79.0	33.0	80.0
6	Male	16.0	94.0	NaN	80.0
7	Female	NaN	90.0	68.0	83.0
8	Male	36.0	50.0	72.0	90.0
9	Male	69.0	90.0	79.0	78.0
10	Female	0.0	83.0	74.0	89.0
11	Female	6.0	87.0	79.0	98.0
12	Male	4.0	87.0	75.0	93.0
13	Male	150.0	43.0	65.0	21.0
14	Male	76.0	75.0	62.0	78.0
15	Female	3.0	46.0	74.0	93.0
16	Male	103.0	NaN	62.0	94.0
17	Male	26.0	78.0	100.0	87.0
18	Male	78.0	95.0	75.0	11.0
19	Female	82.0	82.0	NaN	76.0
20	Female	17.0	82.0	69.0	85.0

	Placement_offer_count	Region
0	8	Nashik
1	6	Pune
2	2	Nashik
3	1	Nashik
4	3	Nashik
5	2	Nashik
6	2	Pune
7	3	Pune
8	2	Nashik
9	1	Nashik
10	2	Pune
11	3	Nashik
12	2	Pune
13	1	Pune
14	1	NaN
15	2	Nashik
16	1	Nashik
17	2	Nashik
18	2	Nashik
19	0	Pune
20	2	Nashik
5.7	5 79.0	

```
In [36]:

IQR = Q3 - Q1
lower_limit = Q1 - 1.5*IQR
upper_limit = Q3 + 1.5*IQR
lower_limit, upper_limit

Out[36]:

(-104.125, 188.875)

In [40]:

col = ["Math_score"]
index_outliers = np.where((df[col] < lower_limit) | (df[col] > upper_limit))
index_outliers

Out[40]:
```

(array([3], dtype=int64), array([0], dtype=int64))

### In [46]:

```
sample_outliers = df[col][(df[col] < lower_limit) | (df[col] > upper_limit)]
sample_outliers
```

### Out[46]:

	Math_score
0	NaN
1	NaN
2	NaN
3	200.0
4	NaN
5	NaN
6	NaN
7	NaN
8	NaN
9	NaN
10	NaN
11	NaN
12	NaN
13	NaN
14	NaN
15	NaN
16	NaN
17	NaN
18	NaN
19	NaN
20	NaN

### In [47]:

```
df1 = df
df[col] = np.where(df[col] < lower_limit, lower_limit, df[col])
df[col] = np.where(df[col] > upper_limit, upper_limit, df[col])
df
```

#### Out[47]:

	Gender	Math_score	Reading_score	Writing_score	Placement_score	Placement_offer_coun
0	Female	15.000	90.0	77.0	88.0	3
1	Female	5.000	82.0	78.0	93.0	ť
2	Female	170.000	100.0	74.0	32.0	2
3	Male	188.875	89.0	32.0	NaN	,
4	NaN	1.000	79.0	67.0	100.0	:
5	Female	56.000	79.0	33.0	80.0	2
6	Male	16.000	94.0	NaN	80.0	2
7	Female	NaN	90.0	68.0	83.0	:
8	Male	36.000	50.0	72.0	90.0	2
9	Male	69.000	90.0	79.0	78.0	,
10	Female	0.000	83.0	74.0	89.0	2
11	Female	6.000	87.0	79.0	98.0	:
12	Male	4.000	87.0	75.0	93.0	2
13	Male	150.000	43.0	65.0	21.0	,
14	Male	76.000	75.0	62.0	78.0	,
15	Female	3.000	46.0	74.0	93.0	2
16	Male	103.000	NaN	62.0	94.0	,
17	Male	26.000	78.0	100.0	87.0	2
18	Male	78.000	95.0	75.0	11.0	2
19	Female	82.000	82.0	NaN	76.0	(
20	Female	17.000	82.0	69.0	85.0	2

```
→
```

### In [49]:

```
ninetieth_percentile = np.nanpercentile(df1["Math_score"], 90)
ninetieth_percentile
```

#### Out[49]:

152.00000000000003

### In [50]:

df1[col] = np.where(df1[col] > upper\_limit, ninetieth\_percentile, df1[col])
df1

### Out[50]:

	Gender	Math_score	Reading_score	Writing_score	Placement_score	Placement_offer_coun
0	Female	15.000	90.0	77.0	88.0	}
1	Female	5.000	82.0	78.0	93.0	•
2	Female	170.000	100.0	74.0	32.0	2
3	Male	188.875	89.0	32.0	NaN	,
4	NaN	1.000	79.0	67.0	100.0	\$
5	Female	56.000	79.0	33.0	80.0	2
6	Male	16.000	94.0	NaN	80.0	2
7	Female	NaN	90.0	68.0	83.0	\$
8	Male	36.000	50.0	72.0	90.0	2
9	Male	69.000	90.0	79.0	78.0	
10	Female	0.000	83.0	74.0	89.0	2
11	Female	6.000	87.0	79.0	98.0	\$
12	Male	4.000	87.0	75.0	93.0	2
13	Male	150.000	43.0	65.0	21.0	
14	Male	76.000	75.0	62.0	78.0	
15	Female	3.000	46.0	74.0	93.0	2
16	Male	103.000	NaN	62.0	94.0	
17	Male	26.000	78.0	100.0	87.0	2
18	Male	78.000	95.0	75.0	11.0	2
19	Female	82.000	82.0	NaN	76.0	(
20	Female	17.000	82.0	69.0	85.0	2
4						<b>•</b>

### In [56]:

median = np.nanmedian(df[col])
median

### Out[56]:

31.0

# In [57]:

```
for i in index_outliers:
    df.at[i, col] = median

df
```

### Out[57]:

	Gender	Math_score	Reading_score	Writing_score	Placement_score	Placement_offer_coun
0	Female	31.0	90.0	77.0	88.0	3
1	Female	5.0	82.0	78.0	93.0	ť
2	Female	170.0	100.0	74.0	32.0	2
3	Male	31.0	89.0	32.0	NaN	•
4	NaN	1.0	79.0	67.0	100.0	:
5	Female	56.0	79.0	33.0	80.0	2
6	Male	16.0	94.0	NaN	80.0	2
7	Female	NaN	90.0	68.0	83.0	:
8	Male	36.0	50.0	72.0	90.0	2
9	Male	69.0	90.0	79.0	78.0	•
10	Female	0.0	83.0	74.0	89.0	2
11	Female	6.0	87.0	79.0	98.0	3
12	Male	4.0	87.0	75.0	93.0	2
13	Male	150.0	43.0	65.0	21.0	•
14	Male	76.0	75.0	62.0	78.0	•
15	Female	3.0	46.0	74.0	93.0	2
16	Male	103.0	NaN	62.0	94.0	
17	Male	26.0	78.0	100.0	87.0	2
18	Male	78.0	95.0	75.0	11.0	2
19	Female	82.0	82.0	NaN	76.0	(
20	Female	17.0	82.0	69.0	85.0	2

 $\blacktriangleleft$ 

### In [59]:

```
import numpy as np
from scipy import stats

df3=pd.read_csv("C:\\Users\\Nancy\\Desktop\\Nancy_Assign2.csv")
df3.head()
df3
```

### Out[59]:

Gender	Math_score	Reading_score	Writing_score	Placement_score	Placement_offer_coun
Female	15.0	90.0	77.0	88.0	3
Female	5.0	82.0	78.0	93.0	ť
Female	170.0	100.0	74.0	32.0	2
Male	200.0	89.0	32.0	NaN	,
NaN	1.0	79.0	67.0	100.0	:
Female	56.0	79.0	33.0	80.0	2
Male	16.0	94.0	NaN	80.0	2
Female	NaN	90.0	68.0	83.0	\$
Male	36.0	50.0	72.0	90.0	2
Male	69.0	90.0	79.0	78.0	•
Female	0.0	83.0	74.0	89.0	2
Female	6.0	87.0	79.0	98.0	\$
Male	4.0	87.0	75.0	93.0	2
Male	150.0	43.0	65.0	21.0	•
Male	76.0	75.0	62.0	78.0	•
Female	3.0	46.0	74.0	93.0	2
Male	103.0	NaN	62.0	94.0	
Male	26.0	78.0	100.0	87.0	2
Male	78.0	95.0	75.0	11.0	2
Female	82.0	82.0	NaN	76.0	(
Female	17.0	82.0	69.0	85.0	2
	Female Female Male NaN Female Male Male Female Male Female Male Female Male Female Male Male Male Male Female Male Female	Female       15.0         Female       5.0         Female       170.0         Male       200.0         NaN       1.0         Female       56.0         Male       16.0         Female       NaN         Male       36.0         Male       69.0         Female       0.0         Female       6.0         Male       4.0         Male       76.0         Female       3.0         Male       103.0         Male       26.0         Male       78.0         Female       82.0	Female       15.0       90.0         Female       5.0       82.0         Female       170.0       100.0         Male       200.0       89.0         NaN       1.0       79.0         Female       56.0       79.0         Male       16.0       94.0         Female       NaN       90.0         Female       NaN       90.0         Female       69.0       90.0         Female       0.0       83.0         Female       6.0       87.0         Male       4.0       87.0         Male       150.0       43.0         Male       76.0       75.0         Female       3.0       46.0         Male       103.0       NaN         Male       26.0       78.0         Male       78.0       95.0         Female       82.0       82.0	Female         15.0         90.0         77.0           Female         5.0         82.0         78.0           Female         170.0         100.0         74.0           Male         200.0         89.0         32.0           NaN         1.0         79.0         67.0           Female         56.0         79.0         33.0           Male         16.0         94.0         NaN           Female         NaN         90.0         68.0           Male         36.0         50.0         72.0           Male         69.0         90.0         79.0           Female         0.0         83.0         74.0           Female         6.0         87.0         79.0           Male         4.0         87.0         75.0           Male         150.0         43.0         65.0           Male         76.0         75.0         62.0           Female         3.0         46.0         74.0           Male         103.0         NaN         62.0           Male         26.0         78.0         100.0           Male         78.0         95.0         75.0	Female         15.0         90.0         77.0         88.0           Female         5.0         82.0         78.0         93.0           Female         170.0         100.0         74.0         32.0           Male         200.0         89.0         32.0         NaN           NaN         1.0         79.0         67.0         100.0           Female         56.0         79.0         33.0         80.0           Male         16.0         94.0         NaN         80.0           Female         NaN         90.0         68.0         83.0           Male         36.0         50.0         72.0         90.0           Male         69.0         90.0         79.0         78.0           Female         0.0         83.0         74.0         89.0           Female         6.0         87.0         79.0         98.0           Male         4.0         87.0         75.0         93.0           Male         150.0         43.0         65.0         21.0           Male         76.0         75.0         62.0         78.0           Female         3.0         46.0         74.0

 $\blacktriangleleft$ 

```
In [70]:
z = np.abs(stats.zscore(df3["Math_score"], nan_policy='omit'))
Z
Out[70]:
      0.690171
0
1
      0.859955
2
      1.941477
3
      2.450828
4
      0.927868
5
      0.005942
      0.673193
6
7
           NaN
      0.333625
8
9
      0.226661
      0.944847
10
11
      0.842976
12
      0.876933
13
      1.601910
14
      0.345510
15
      0.893911
16
      0.803926
17
      0.503409
      0.379467
18
19
      0.447380
20
      0.656214
Name: Math_score, dtype: float64
In [71]:
threshold = 0.18
sample_outliers = np.where(z < threshold)</pre>
sample_outliers
Out[71]:
(array([5], dtype=int64),)
In [74]:
upper threshold = 1.4
lower threshold = 0.18
sample_outliers = np.where((z < lower_threshold) | (z > upper_threshold))
sample_outliers
Out[74]:
```

(array([ 2, 3, 5, 13], dtype=int64),)

# In [75]:

```
new_df1 = df3
for i in sample_outliers:
    new_df1.drop(i, inplace=True)
new_df1
```

# Out[75]:

	Gender	Math_score	Reading_score	Writing_score	Placement_score	Placement_offer_coun
0	Female	15.0	90.0	77.0	88.0	{
1	Female	5.0	82.0	78.0	93.0	(
4	NaN	1.0	79.0	67.0	100.0	3
6	Male	16.0	94.0	NaN	80.0	2
7	Female	NaN	90.0	68.0	83.0	:
8	Male	36.0	50.0	72.0	90.0	2
9	Male	69.0	90.0	79.0	78.0	,
10	Female	0.0	83.0	74.0	89.0	2
11	Female	6.0	87.0	79.0	98.0	:
12	Male	4.0	87.0	75.0	93.0	2
14	Male	76.0	75.0	62.0	78.0	,
15	Female	3.0	46.0	74.0	93.0	2
16	Male	103.0	NaN	62.0	94.0	,
17	Male	26.0	78.0	100.0	87.0	2
18	Male	78.0	95.0	75.0	11.0	2
19	Female	82.0	82.0	NaN	76.0	(
20	Female	17.0	82.0	69.0	85.0	2
4						<b>•</b>

In [ ]: