

```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
```

```
In [3]: df=pd.read_csv("/home/mca01/Downloads/StudentsPerformance.csv")
```

```
In [4]: df.head(15)
```

```
Out[4]:
```

	gender	race/ethnicity	parental level of education	lunch	test preparation course	math score	readin score
0	female	group B	bachelor's degree	standard	none	72	7
1	female	group C	some college	standard	completed	69	9
2	female	group B	master's degree	standard	none	90	9
3	male	group A	associate's degree	free/reduced	none	47	5
4	male	group C	some college	standard	none	76	7
5	female	group B	associate's degree	standard	none	71	8
6	female	group B	some college	standard	completed	88	9
7	male	group B	some college	free/reduced	none	40	4
8	male	group D	high school	free/reduced	completed	64	6
9	female	group B	high school	free/reduced	none	38	6
10	male	group C	associate's degree	standard	none	58	5
11	male	group D	associate's degree	standard	none	40	5
12	female	group B	high school	standard	none	65	8
13	male	group A	some college	standard	completed	78	7
14	female	group A	master's degree	standard	none	50	5

```
In [5]: df.shape
```

Out[5]: (1000, 8)

```
In [6]: df.dtypes.value_counts()
```

```
Out[6]: object    5
        int64     3
        dtype: int64
```

```
In [7]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1000 entries, 0 to 999
Data columns (total 8 columns):
#   Column                                          Non-Null Count  Dtype
---  -
0   gender                                          1000 non-null   object
1   race/ethnicity                                1000 non-null   object
2   parental level of education                    1000 non-null   object
3   lunch                                           1000 non-null   object
4   test preparation course                        1000 non-null   object
5   math score                                     1000 non-null   int64
6   reading score                                 1000 non-null   int64
7   writing score                                  1000 non-null   int64
dtypes: int64(3), object(5)
memory usage: 62.6+ KB
```

```
In [8]: df.describe()
```

```
Out[8]:
```

	math score	reading score	writing score
count	1000.00000	1000.000000	1000.000000
mean	66.08900	69.169000	68.054000
std	15.16308	14.600192	15.195657
min	0.00000	17.000000	10.000000
25%	57.00000	59.000000	57.750000
50%	66.00000	70.000000	69.000000
75%	77.00000	79.000000	79.000000
max	100.00000	100.000000	100.000000

Handle the Missing Value

```
In [9]: df.isnull().sum()
```

```
Out[9]: gender          0
        race/ethnicity  0
        parental level of education  0
        lunch           0
        test preparation course  0
        math score      0
        reading score   0
        writing score    0
        dtype: int64
```

Making list of columns having missing value

```
In [10]: data = df
        coln=[]
        miss=[]
        coln.extend(data.columns)
        for i in coln:
            t=data[i].isnull

            if t!=0:
                miss.append(i)
            else:
                continue
        print(miss)
```

```
['gender', 'race/ethnicity', 'parental level of education', 'lunch', 'test p
reparation course', 'math score', 'reading score', 'writing score']
```

```
In [11]: pd.options.mode.chained_assignment=None
        for j in miss:
            q=data[j].dtypes
            if(q=='int64' or q=='float64'):
                f=data[j]
                for k in range(data.shape[0]):
                    if(f[k]<0 or f[k]>100):
                        f[k]=(np.nan)
            else:
                data.fillna(method='bfill')

        data.head(20)
```

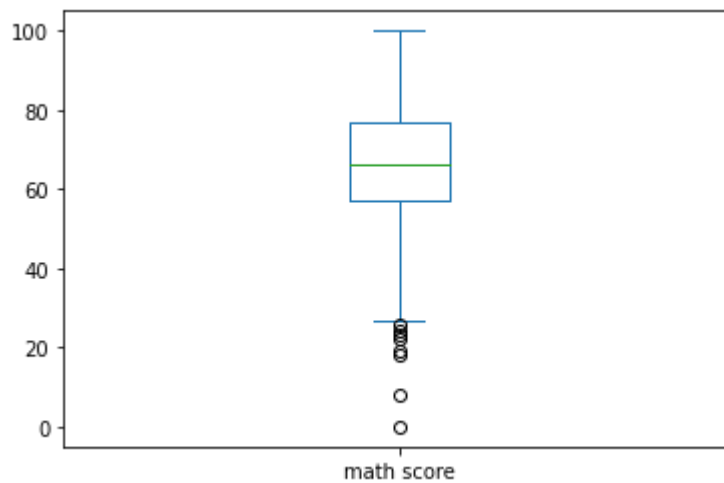
Out[11]:

	gender	race/ethnicity	parental level of education	lunch	test preparation course	math score	readin score
0	female	group B	bachelor's degree	standard	none	72	7
1	female	group C	some college	standard	completed	69	9
2	female	group B	master's degree	standard	none	90	9
3	male	group A	associate's degree	free/reduced	none	47	5
4	male	group C	some college	standard	none	76	7
5	female	group B	associate's degree	standard	none	71	8
6	female	group B	some college	standard	completed	88	9
7	male	group B	some college	free/reduced	none	40	4
8	male	group D	high school	free/reduced	completed	64	6
9	female	group B	high school	free/reduced	none	38	6
10	male	group C	associate's degree	standard	none	58	5
11	male	group D	associate's degree	standard	none	40	5
12	female	group B	high school	standard	none	65	8
13	male	group A	some college	standard	completed	78	7
14	female	group A	master's degree	standard	none	50	5
15	female	group C	some high school	standard	none	69	7
16	male	group C	high school	standard	none	88	8
17	female	group B	some high school	free/reduced	none	18	3
18	male	group C	master's degree	free/reduced	completed	46	4
19	female	group C	associate's degree	free/reduced	none	54	5

Review of Parents

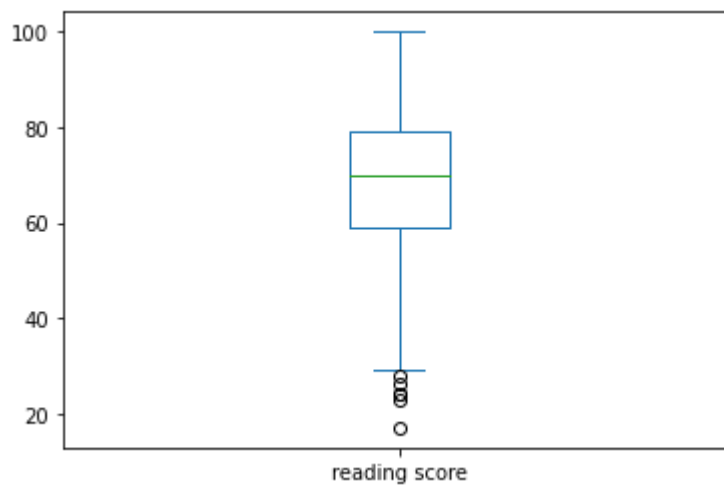
```
In [12]: data['math score'].plot(kind='box')
```

```
Out[12]: <AxesSubplot:>
```



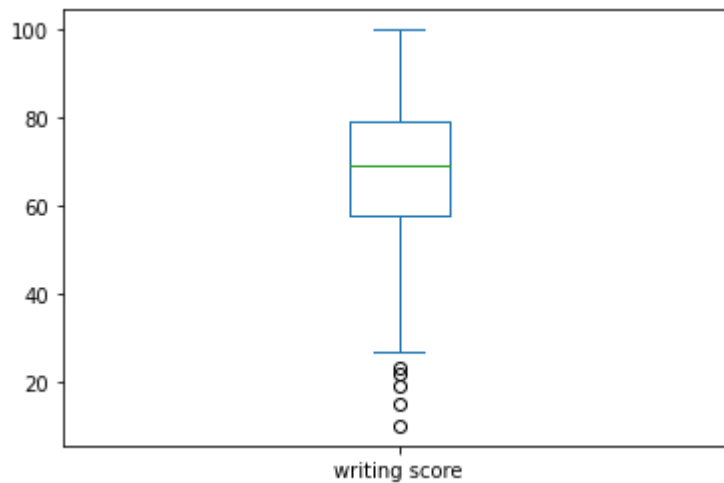
```
In [13]: data['reading score'].plot(kind='box')
```

```
Out[13]: <AxesSubplot:>
```



```
In [14]: data['writing score'].plot(kind='box')
```

```
Out[14]: <AxesSubplot:>
```



```
In [15]: data.head()
```

```
Out[15]:
```

	gender	race/ethnicity	parental level of education	lunch	test preparation course	math score	reading score
0	female	group B	bachelor's degree	standard	none	72	72
1	female	group C	some college	standard	completed	69	90
2	female	group B	master's degree	standard	none	90	95
3	male	group A	associate's degree	free/reduced	none	47	57
4	male	group C	some college	standard	none	76	78

Outliers Removal

```
In [16]: q1=data['math score'].quantile(0.25)
q3=data['math score'].quantile(0.75)
iqr = q3-q1

lowerlimit=q1 - 1.5*iqr
upperlimit =q3 + 1.5*iqr

print("Q1",q1, "\nQ3:" , q3,"\nIQR:",iqr, "\nLOWER LIMIT",lowerlimit,"\nUPPER LIMIT",upperlimit)

Q1 57.0
Q3: 77.0
IQR: 20.0
LOWER LIMIT 27.0
UPPER LIMIT 107.0
```

```
In [17]: data[(data['math score']<lowerlimit)|(data['math score']>upperlimit)]
```

Out[17]:

	gender	race/ethnicity	parental level of education	lunch	test preparation course	math score	reading score
17	female	group B	some high school	free/reduced	none	18	
59	female	group C	some high school	free/reduced	none	0	
145	female	group C	some college	free/reduced	none	22	
338	female	group B	some high school	free/reduced	none	24	
466	female	group D	associate's degree	free/reduced	none	26	
787	female	group B	some college	standard	none	19	
842	female	group B	high school	free/reduced	completed	23	
980	female	group B	high school	free/reduced	none	8	

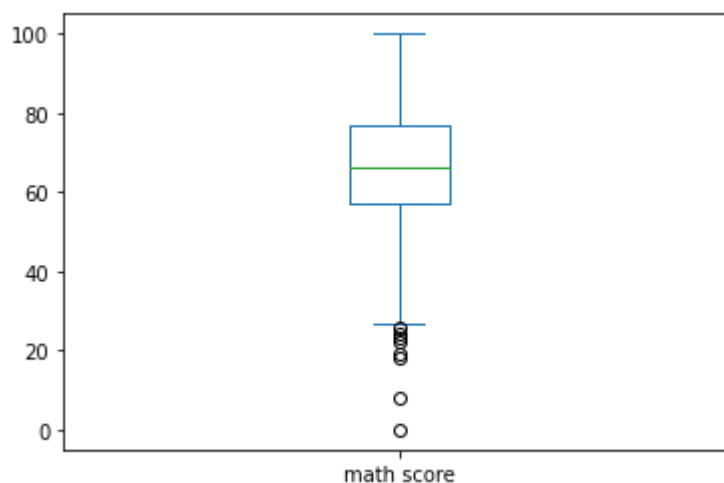
```
In [18]: data[(data['math score']<lowerlimit)&(data['math score']>upperlimit)]
```

Out[18]:

	gender	race/ethnicity	parental level of education	lunch	test preparation course	math score	reading score	writing score
--	--------	----------------	-----------------------------------	-------	-------------------------------	---------------	------------------	------------------

```
In [19]: data['math score'].plot(kind='box')
```

Out[19]: <AxesSubplot:>



Zscore Scaling

In [20]: data

Out[20]:

	gender	race/ethnicity	parental level of education	lunch	test preparation course	math score	reading score
0	female	group B	bachelor's degree	standard	none	72	
1	female	group C	some college	standard	completed	69	
2	female	group B	master's degree	standard	none	90	
3	male	group A	associate's degree	free/reduced	none	47	
4	male	group C	some college	standard	none	76	
...
995	female	group E	master's degree	standard	completed	88	
996	male	group C	high school	free/reduced	none	62	
997	female	group C	high school	free/reduced	completed	59	
998	female	group D	some college	standard	completed	68	
999	female	group D	some college	free/reduced	none	77	

1000 rows × 8 columns

```
In [21]: new_data=data
from scipy import stats
```

```
In [22]: columns=['math score','reading score','writing score']
new_data[columns]= stats.zscore(new_data[columns])
new_data
```


Out[22]:

	gender	race/ethnicity	parental level of education	lunch	test preparation course	math score	
0	female	group B	bachelor's degree	standard	none	0.390024	0
1	female	group C	some college	standard	completed	0.192076	1
2	female	group B	master's degree	standard	none	1.577711	1
3	male	group A	associate's degree	free/reduced	none	-1.259543	-0
4	male	group C	some college	standard	none	0.653954	0
...
995	female	group E	master's degree	standard	completed	1.445746	2
996	male	group C	high school	free/reduced	none	-0.269803	-0
997	female	group C	high school	free/reduced	completed	-0.467751	0
998	female	group D	some college	standard	completed	0.126093	0
999	female	group D	some college	free/reduced	none	0.719937	1

1000 rows × 8 columns

MinMax scaling

```
In [23]: new_data1=data
```

```
In [30]: from sklearn.preprocessing import MinMaxScaler  
scaler = MinMaxScaler()
```

```
In [31]: col=['math score','reading score','writing score']  
scaler.fit(new_data1[col])  
new_data1[col]=scaler.transform(new_data1[col])
```

```
In [32]: new_data1
```

Out[32]:

	gender	race/ethnicity	parental level of education	lunch	test preparation course	math score	read sc
0	female	group B	bachelor's degree	standard	none	0.72	0.6620
1	female	group C	some college	standard	completed	0.69	0.879
2	female	group B	master's degree	standard	none	0.90	0.939
3	male	group A	associate's degree	free/reduced	none	0.47	0.481
4	male	group C	some college	standard	none	0.76	0.734
...
995	female	group E	master's degree	standard	completed	0.88	0.987
996	male	group C	high school	free/reduced	none	0.62	0.457
997	female	group C	high school	free/reduced	completed	0.59	0.650
998	female	group D	some college	standard	completed	0.68	0.734
999	female	group D	some college	free/reduced	none	0.77	0.831

1000 rows × 8 columns

In []: