

```
In [61]: import pandas as pd
import numpy as np
from scipy import stats
from scipy.stats import norm
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
import warnings
warnings.filterwarnings('ignore')
```

## Q7) Calculate Mean, Median, Mode, Variance, Standard Deviation, Range & comment about the values / draw inferences, for the given dataset

- -For Points,Score,Weigh>
- Find Mean, Median, Mode, Variance, Standard Deviation, and Range and also Comment about the values/ Draw some inferences.

```
In [62]: df=pd.read_csv('Q7.csv')
df
```

Out[62]:

	Unnamed: 0	Points	Score	Weigh
0	Mazda RX4	3.90	2.620	16.46
1	Mazda RX4 Wag	3.90	2.875	17.02
2	Datsun 710	3.85	2.320	18.61
3	Hornet 4 Drive	3.08	3.215	19.44
4	Hornet Sportabout	3.15	3.440	17.02
5	Valiant	2.76	3.460	20.22
6	Duster 360	3.21	3.570	15.84
7	Merc 240D	3.69	3.190	20.00
8	Merc 230	3.92	3.150	22.90
9	Merc 280	3.92	3.440	18.30
10	Merc 280C	3.92	3.440	18.90
11	Merc 450SE	3.07	4.070	17.40
12	Merc 450SL	3.07	3.730	17.60
13	Merc 450SLC	3.07	3.780	18.00
14	Cadillac Fleetwood	2.93	5.250	17.98
15	Lincoln Continental	3.00	5.424	17.82
16	Chrysler Imperial	3.23	5.345	17.42
17	Fiat 128	4.08	2.200	19.47
18	Honda Civic	4.93	1.615	18.52
19	Toyota Corolla	4.22	1.835	19.90
20	Toyota Corona	3.70	2.465	20.01
21	Dodge Challenger	2.76	3.520	16.87
22	AMC Javelin	3.15	3.435	17.30
23	Camaro Z28	3.73	3.840	15.41

	Unnamed: 0	Points	Score	Weigh
24	Pontiac Firebird	3.08	3.845	17.05
25	Fiat X1-9	4.08	1.935	18.90
26	Porsche 914-2	4.43	2.140	16.70
27	Lotus Europa	3.77	1.513	16.90
28	Ford Pantera L	4.22	3.170	14.50
29	Ferrari Dino	3.62	2.770	15.50
30	Maserati Bora	3.54	3.570	14.60
31	Volvo 142E	4.11	2.780	18.60

In [63]: `df.describe()`

Out[63]:

	Points	Score	Weigh
<b>count</b>	32.000000	32.000000	32.000000
<b>mean</b>	3.596563	3.217250	17.848750
<b>std</b>	0.534679	0.978457	1.786943
<b>min</b>	2.760000	1.513000	14.500000
<b>25%</b>	3.080000	2.581250	16.892500
<b>50%</b>	3.695000	3.325000	17.710000
<b>75%</b>	3.920000	3.610000	18.900000
<b>max</b>	4.930000	5.424000	22.900000

## Mean

In [64]: `df.mean()`

Out[64]:

```
Points    3.596563
Score     3.217250
Weigh     17.848750
dtype: float64
```

## Variance

```
In [65]: df.var()
```

```
Out[65]: Points    0.285881  
Score    0.957379  
Weigh    3.193166  
dtype: float64
```

## Standard Deviation

```
In [66]: df.std()
```

```
Out[66]: Points    0.534679  
Score    0.978457  
Weigh    1.786943  
dtype: float64
```

## Median

```
In [67]: df.median()
```

```
Out[67]: Points    3.695  
Score    3.325  
Weigh    17.710  
dtype: float64
```

```
In [68]: df.min()
```

```
Out[68]: Unnamed: 0    AMC Javelin  
Points    2.76  
Score    1.513  
Weigh    14.5  
dtype: object
```

```
In [69]: df.max()
```

```
Out[69]: Unnamed: 0      Volvo 142E  
Points          4.93  
Score           5.424  
Weigh           22.9  
dtype: object
```

## mode

```
In [70]: df.Points.mode()
```

```
Out[70]: 0      3.07  
1      3.92  
Name: Points, dtype: float64
```

```
In [71]: df.Score.mode()
```

```
Out[71]: 0      3.44  
Name: Score, dtype: float64
```

```
In [72]: df.Weigh.mode()
```

```
Out[72]: 0      17.02  
1      18.90  
Name: Weigh, dtype: float64
```

## Range

```
In [73]: df_range=df.Points.max()-df.Points.min()  
df_range
```

```
Out[73]: 2.17
```

```
In [74]: score_range=df.Score.max()-df.Score.min()  
score_range
```

```
Out[74]: 3.9110000000000005
```

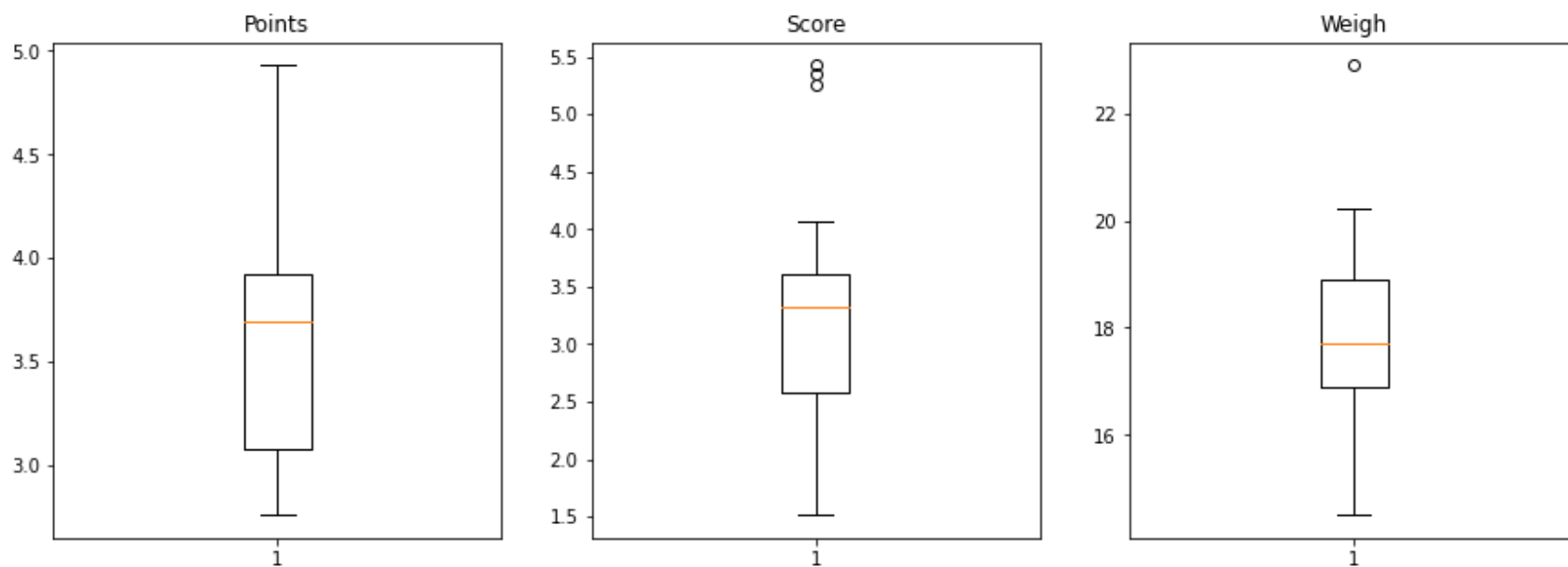
```
In [75]: weigh_range=df.Weigh.max()-df.Weigh.min()  
weigh_range
```

```
Out[75]: 8.399999999999999
```

## infrances

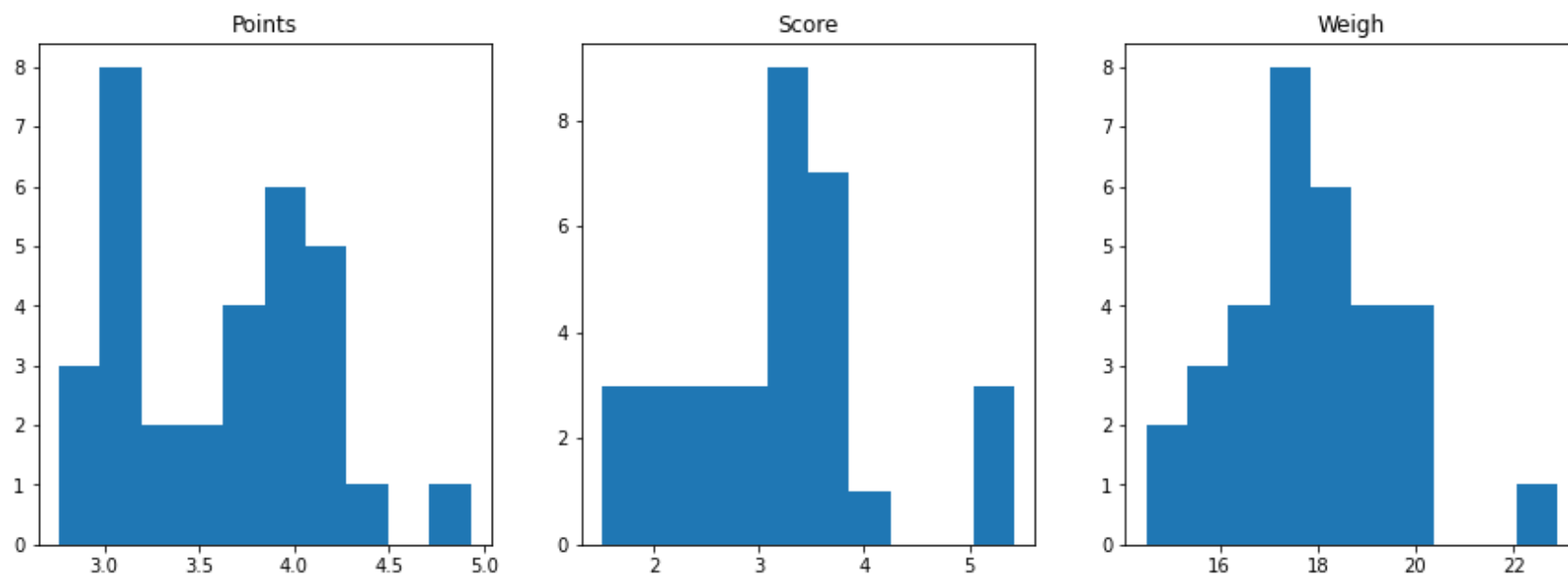
### Boxplot

```
In [76]: Fig=plt.subplots(figsize=(15,5))
plt.subplot(1,3,1)
plt.boxplot(df.Points)
plt.title('Points')
plt.subplot(1,3,2)
plt.boxplot(df.Score)
plt.title('Score')
plt.subplot(1,3,3)
plt.boxplot(df.Weigh)
plt.title('Weigh')
plt.show()
```



## Histogram

```
In [77]: Fig=plt.subplots(figsize=(15,5))
plt.subplot(1,3,1)
plt.hist(df.Points)
plt.title('Points')
plt.subplot(1,3,2)
plt.hist(df.Score)
plt.title('Score')
plt.subplot(1,3,3)
plt.hist(df.Weigh)
plt.title('Weigh')
plt.show()
```





## **Q9) Calculate Skewness, Kurtosis & draw inferences on the following data**

- Cars speed and distance
- Use Q9\_a.csv
- SP and Weight(WT)

```
In [78]: Q_9=pd.read_csv('Q9_a.csv')
```

In [79]: Q\_9

Out[79]:

	Index	speed	dist
0	1	4	2
1	2	4	10
2	3	7	4
3	4	7	22
4	5	8	16
5	6	9	10
6	7	10	18
7	8	10	26
8	9	10	34
9	10	11	17
10	11	11	28
11	12	12	14
12	13	12	20
13	14	12	24
14	15	12	28
15	16	13	26
16	17	13	34
17	18	13	34
18	19	13	46
19	20	14	26
20	21	14	36
21	22	14	60
22	23	14	80
23	24	15	20
24	25	15	26

	Index	speed	dist
<b>25</b>	26	15	54
<b>26</b>	27	16	32
<b>27</b>	28	16	40
<b>28</b>	29	17	32
<b>29</b>	30	17	40
<b>30</b>	31	17	50
<b>31</b>	32	18	42
<b>32</b>	33	18	56
<b>33</b>	34	18	76
<b>34</b>	35	18	84
<b>35</b>	36	19	36
<b>36</b>	37	19	46
<b>37</b>	38	19	68
<b>38</b>	39	20	32
<b>39</b>	40	20	48
<b>40</b>	41	20	52
<b>41</b>	42	20	56
<b>42</b>	43	20	64
<b>43</b>	44	22	66
<b>44</b>	45	23	54
<b>45</b>	46	24	70
<b>46</b>	47	24	92
<b>47</b>	48	24	93
<b>48</b>	49	24	120
<b>49</b>	50	25	85

```
In [80]: Q_9.iloc[:,1:]
```

```
Out[80]:
```

	speed	dist
0	4	2
1	4	10
2	7	4
3	7	22
4	8	16
5	9	10
6	10	18
7	10	26
8	10	34
9	11	17
10	11	28
11	12	14
12	12	20
13	12	24
14	12	28
15	13	26
16	13	34
17	13	34
18	13	46
19	14	26
20	14	36
21	14	60
22	14	80
23	15	20
24	15	26

	speed	dist
25	15	54
26	16	32
27	16	40
28	17	32
29	17	40
30	17	50
31	18	42
32	18	56
33	18	76
34	18	84
35	19	36
36	19	46
37	19	68
38	20	32
39	20	48
40	20	52
41	20	56
42	20	64
43	22	66
44	23	54
45	24	70
46	24	92
47	24	93
48	24	120
49	25	85

## Skewnes

```
In [81]: Q_9.skew()
```

```
Out[81]: Index      0.000000  
         speed     -0.117510  
         dist      0.806895  
         dtype: float64
```

## Kurtosis

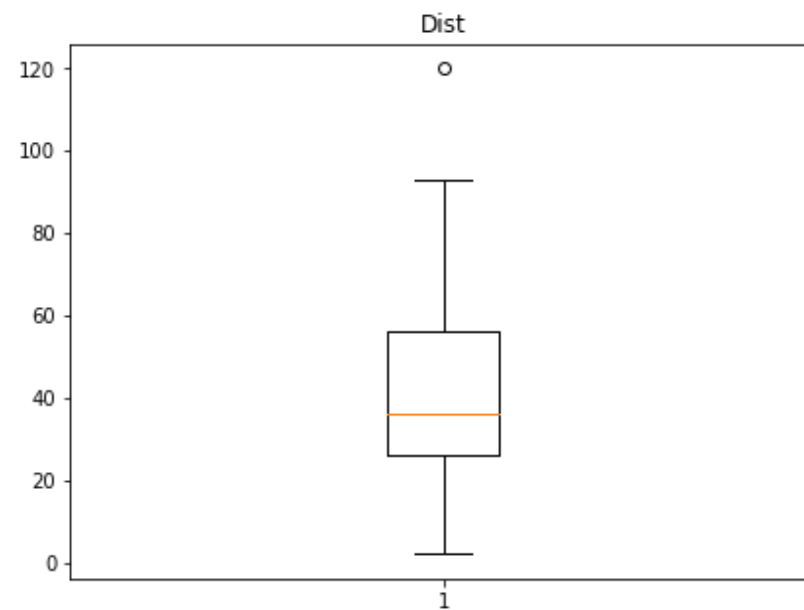
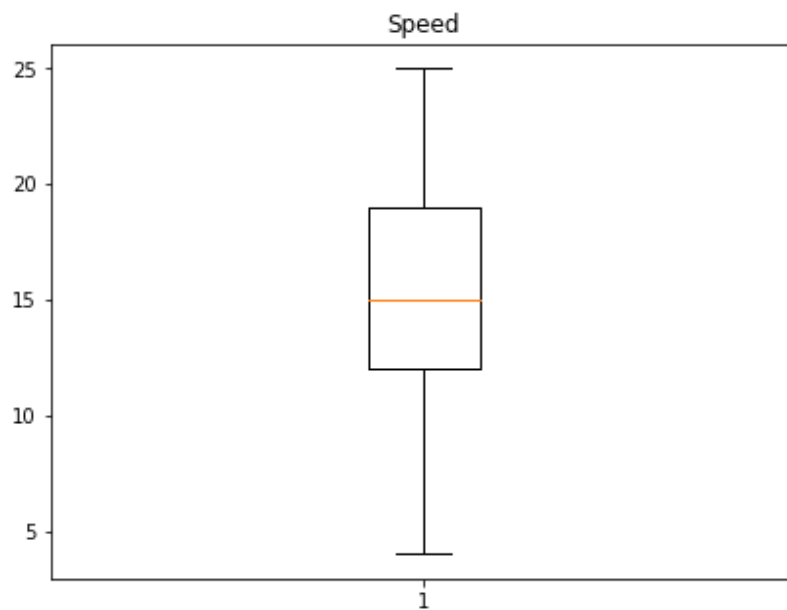
```
In [82]: Q_9.kurt()
```

```
Out[82]: Index      -1.200000  
         speed     -0.508994  
         dist      0.405053  
         dtype: float64
```

## inferences

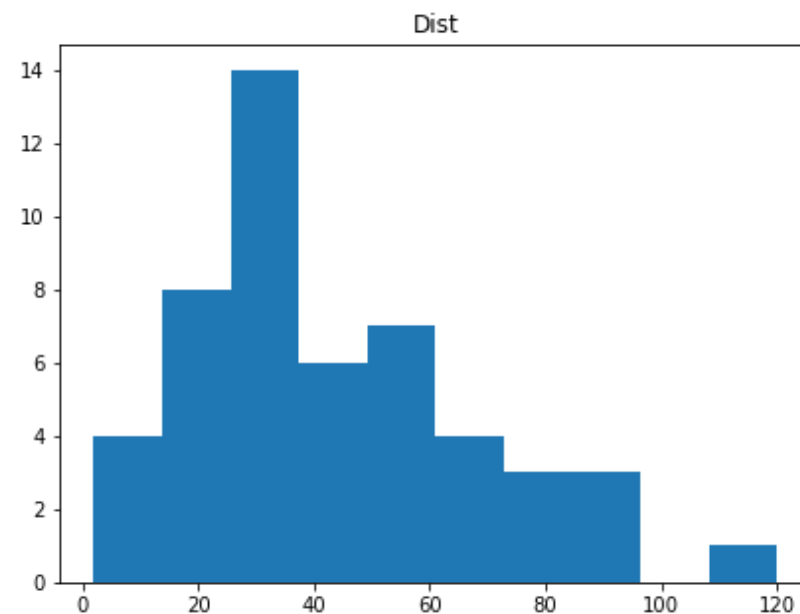
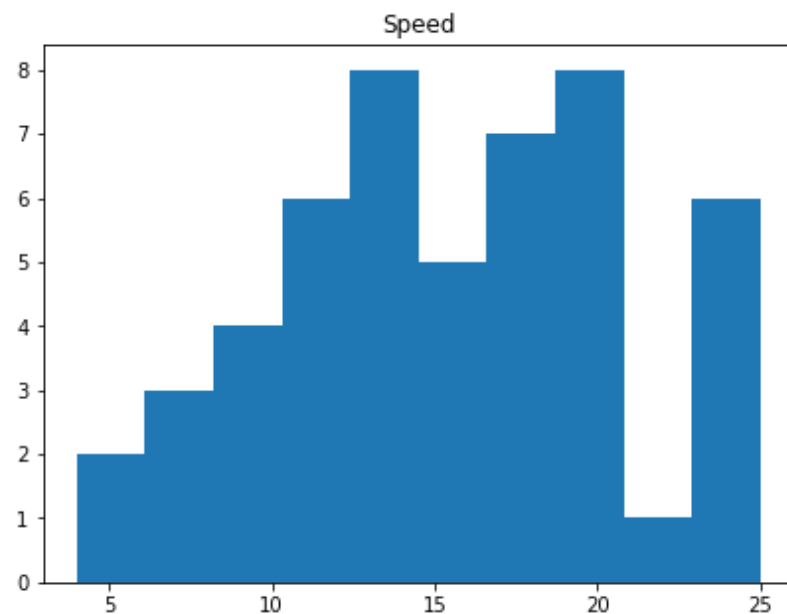
## Boxplot

```
In [83]: Fig=plt.subplots(figsize=(15,5))
plt.subplot(1,2,1)
plt.boxplot(Q_9.speed)
plt.title('Speed')
plt.subplot(1,2,2)
plt.boxplot(Q_9.dist)
plt.title('Dist')
plt.show()
```



## Histogram

```
In [84]: Fig=plt.subplots(figsize=(15,5))
plt.subplot(1,2,1)
plt.hist(Q_9.speed)
plt.title('Speed')
plt.subplot(1,2,2)
plt.hist(Q_9.dist)
plt.title('Dist')
plt.show()
```



=====

**Q\_9(b)**



# Import Dataset

```
In [85]: Q_9b = pd.read_csv('Q9_b.csv')  
Q_9b
```

Out[85]:

	Unnamed: 0	SP	WT
0	1	104.185353	28.762059
1	2	105.461264	30.466833
2	3	105.461264	30.193597
3	4	113.461264	30.632114
4	5	104.461264	29.889149
...	...	...	...
76	77	169.598513	16.132947
77	78	150.576579	37.923113
78	79	151.598513	15.769625
79	80	167.944460	39.423099
80	81	139.840817	34.948615

81 rows × 3 columns

## Removing "Unnamed:0" Column

```
In [86]: Q_9b = Q_9b.iloc[:,1:]  
Q_9b
```

Out[86]:

	SP	WT
0	104.185353	28.762059
1	105.461264	30.466833
2	105.461264	30.193597
3	113.461264	30.632114
4	104.461264	29.889149
...	...	...
76	169.598513	16.132947
77	150.576579	37.923113
78	151.598513	15.769625
79	167.944460	39.423099
80	139.840817	34.948615

81 rows × 2 columns

## Skewness

```
In [87]: Q_9b.skew()
```

Out[87]: SP 1.611450  
WT -0.614753  
dtype: float64

## Kurtosis

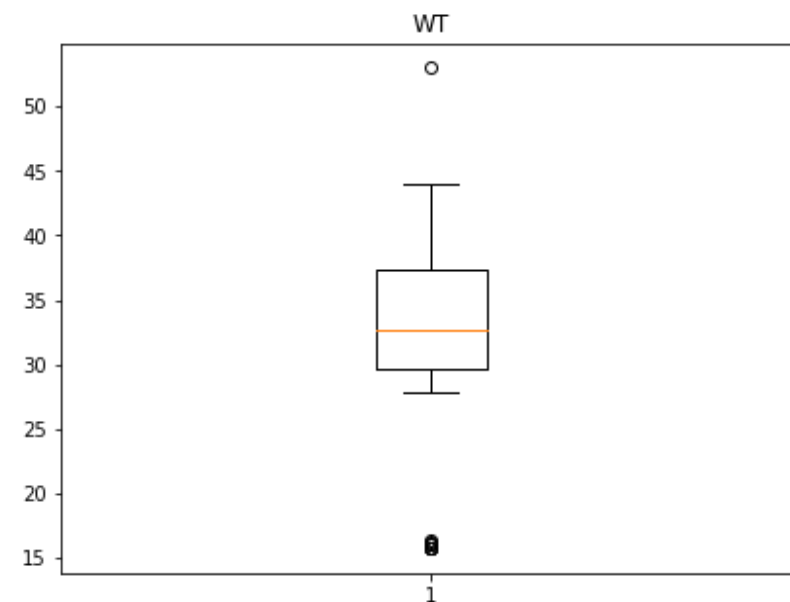
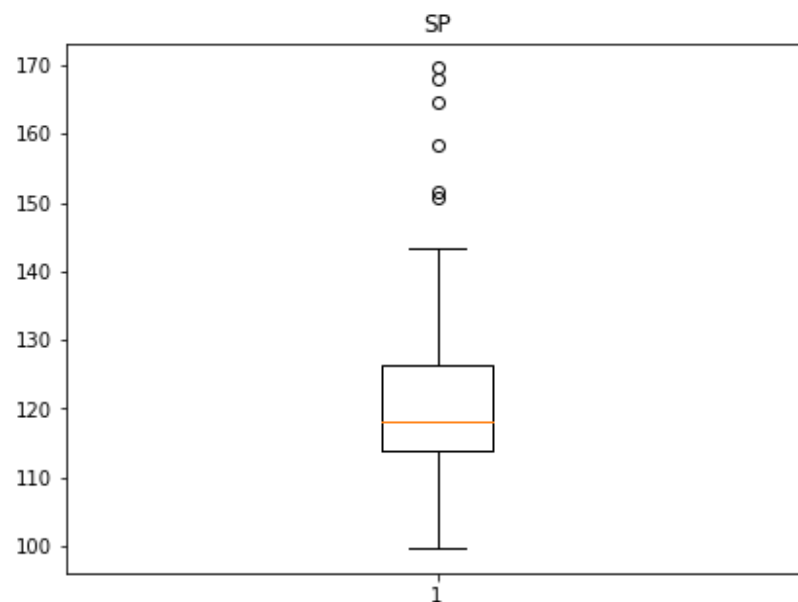
```
In [88]: Q_9b.kurt()
```

```
Out[88]: SP      2.977329  
         WT      0.950291  
         dtype: float64
```

## Inferences

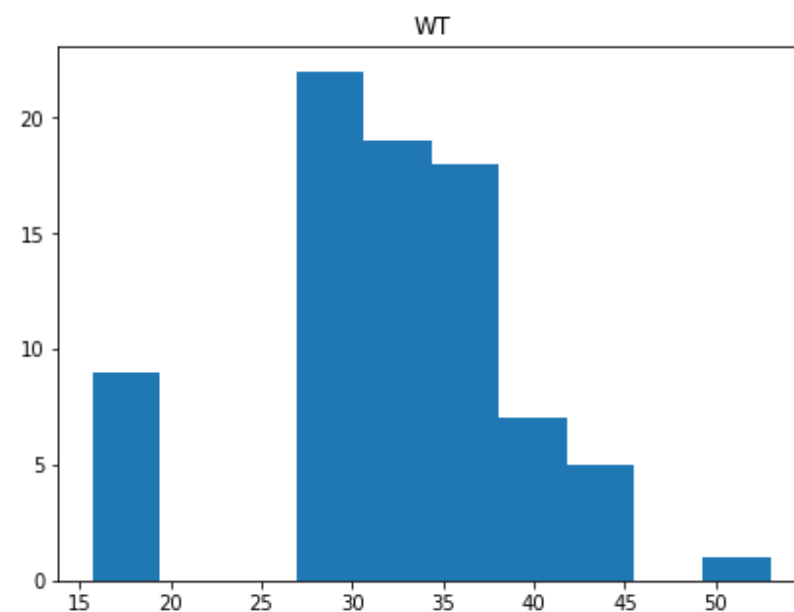
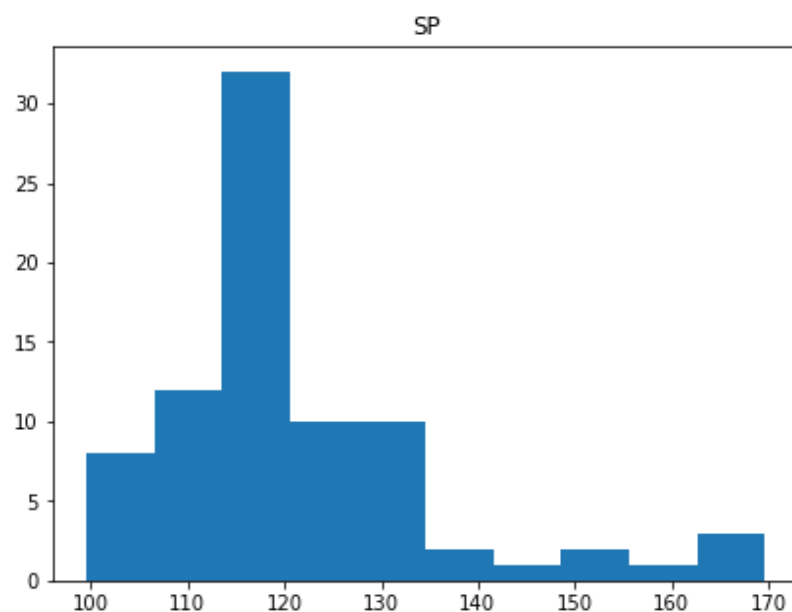
### Boxplot

```
In [89]: Fig=plt.subplots(figsize=(15,5))
plt.subplot(1,2,1)
plt.boxplot(Q_9b.SP)
plt.title('SP')
plt.subplot(1,2,2)
plt.boxplot(Q_9b.WT)
plt.title('WT')
plt.show()
```



## Histogram

```
In [90]: Fig=plt.subplots(figsize=(15,5))
plt.subplot(1,2,1)
plt.hist(Q_9b.SP)
plt.title('SP')
plt.subplot(1,2,2)
plt.hist(Q_9b.WT)
plt.title('WT')
plt.show()
```



=====

**Q\_11. Suppose we want to estimate the average weight of an adult male in Mexico. We draw a random sample of 2,000 men from a population of 3,000,000 men and weigh them. We find that the average person in our sample weighs 200 pounds, and the standard deviation of the sample is 30 pounds. Calculate 94%,98%,96% confidence interval?**

**Avg. weight of adult in mexico with 94% ci**

```
In [91]: stats.norm.interval(0.94,200,30/(2000**0.5))
```

```
Out[91]: (198.738325292158, 201.261674707842)
```

**Avg. weight of adult in mexico with 98% ci**

```
In [92]: stats.norm.interval(0.98,200,30/(2000**0.5))
```

```
Out[92]: (198.43943840429978, 201.56056159570022)
```

**Avg. weight of adult in mexico with 96% ci**

```
In [93]: stats.norm.interval(0.96,200,30/(2000**0.5))
```

```
Out[93]: (198.62230334813333, 201.37769665186667)
```

=====

◀ ▶

## Q\_12. Below are the scores obtained by a student in tests

34,36,36,38,38,39,39,40,40,41,41,41,41,42,42,45,49,56

1) Find mean, median, variance, standard deviation.

2) What can we say about the student marks?

```
In [94]: ab = pd.Series([34,36,36,38,38,39,39,40,40,41,41,41,41,42,42,45,49,56])
```

### mean

```
In [95]: ab.mean()
```

```
Out[95]: 41.0
```

### Median

```
In [96]: ab.median()
```

```
Out[96]: 40.5
```

### Variance

```
In [97]: ab.var()
```

```
Out[97]: 25.529411764705884
```

### Standard Deviation

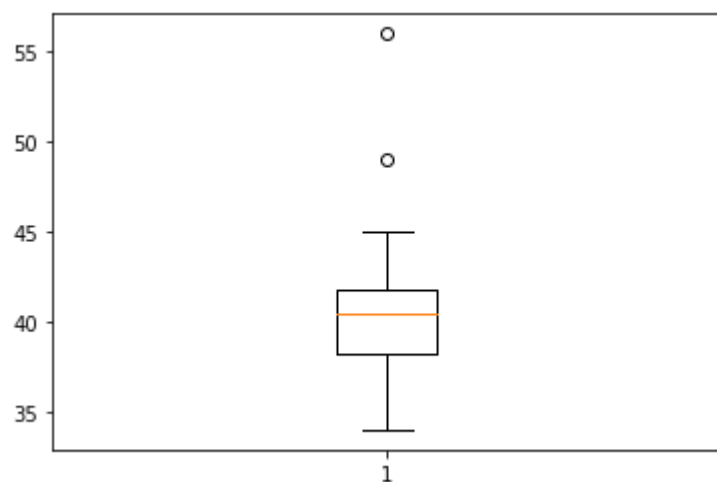
```
In [98]: ab.std()
```

```
Out[98]: 5.05266382858645
```

## Inferences

### Boxplot

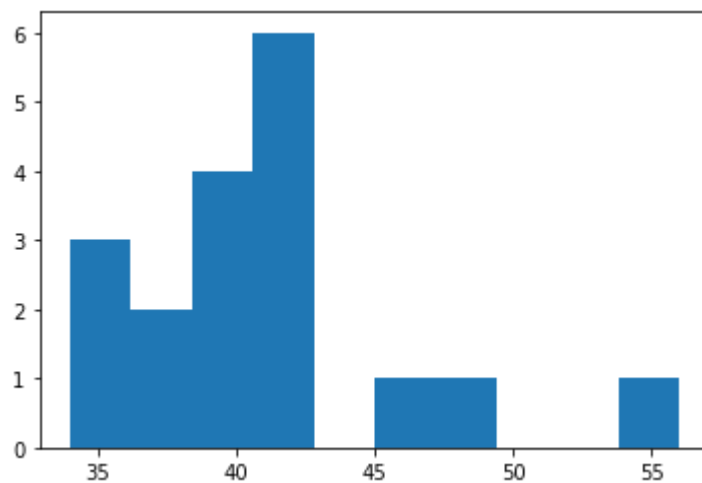
```
In [99]: plt.boxplot(ab)  
plt.show()
```



### Histogram



```
In [100]: plt.hist(ab,label='ab')  
plt.show()
```



## Q\_20.Calculate probability from the given dataset for the below cases

Data \_set: Cars.csv

Calculate the probability of MPG of Cars for the below cases.

MPG <- Cars\$MPG a.  $P(\text{MPG} > 38)$

b.  $P(\text{MPG} < 40)$

c.  $P(20 < \text{MPG} < 50)$

## import Dataset

```
In [105]: Cars = pd.read_csv('Cars (1).csv')
Cars
```

Out[105]:

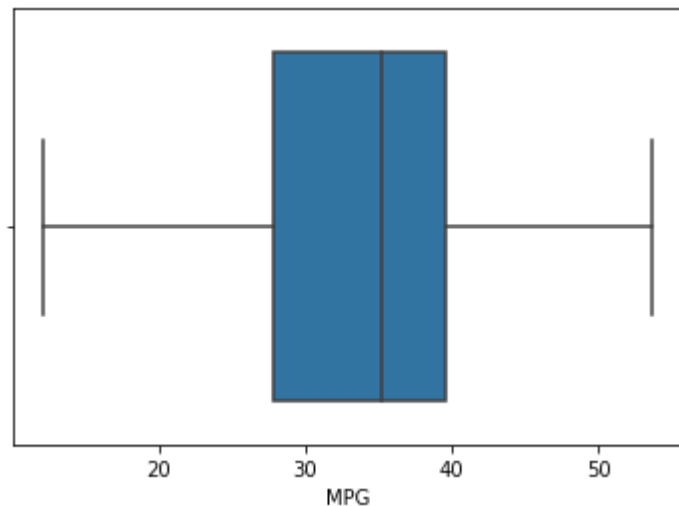
	HP	MPG	VOL	SP	WT
0	49	53.700681	89	104.185353	28.762059
1	55	50.013401	92	105.461264	30.466833
2	55	50.013401	92	105.461264	30.193597
3	70	45.696322	92	113.461264	30.632114
4	53	50.504232	92	104.461264	29.889149
...	...	...	...	...	...
76	322	36.900000	50	169.598513	16.132947
77	238	19.197888	115	150.576579	37.923113
78	263	34.000000	50	151.598513	15.769625
79	295	19.833733	119	167.944460	39.423099
80	236	12.101263	107	139.840817	34.948615

81 rows × 5 columns

## Boxplot

```
In [107]: sns.boxplot(Cars.MPG)
```

```
Out[107]: <AxesSubplot:xlabel='MPG'>
```



## A. $P(\text{MPG} > 38)$

```
In [108]: 1-stats.norm.cdf(38,Cars.MPG.mean(),Cars.MPG.std())
```

```
Out[108]: 0.3475939251582705
```

## B. $P(\text{MPG} < 40)$

```
In [109]: 1-stats.norm.cdf(40,Cars.MPG.mean(),Cars.MPG.std())
```

```
Out[109]: 0.27065012378483844
```

## C. P (20<MPG<50)

```
In [110]: stats.norm.cdf(0.50,Cars.MPG.mean(),Cars.MPG.std())-stats.norm.cdf(0.20,Cars.MPG.mean(),Cars.MPG.std())
```

```
Out[110]: 1.2430968797327613e-05
```



## Q\_21.Check whether the data follows normal distribution¶

- a) Check whether the MPG of Cars follows Normal Distribution
- b) Check Whether the Adipose Tissue (AT) and Waist Circumference(Waist) from wc-at data set follows Normal Distribution

### Q 21(a)

Import Important Libraries and Warnings

## import Dataset

```
In [111]: Cars = pd.read_csv('Cars (1).csv')
Cars
```

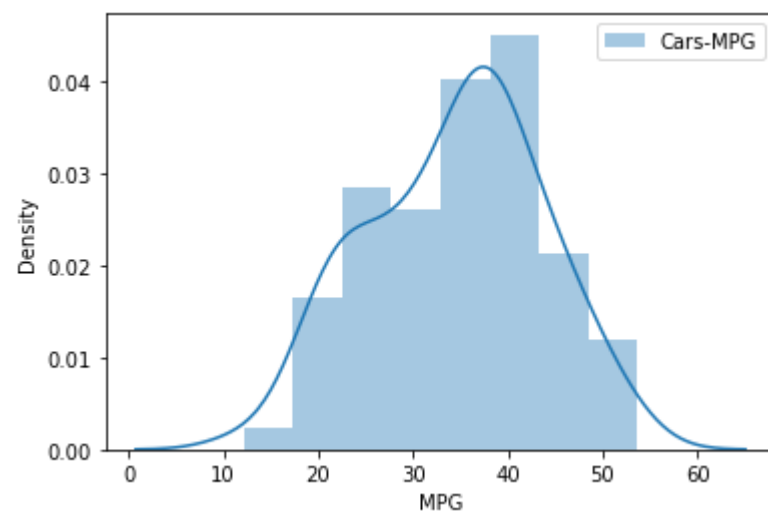
Out[111]:

	HP	MPG	VOL	SP	WT
0	49	53.700681	89	104.185353	28.762059
1	55	50.013401	92	105.461264	30.466833
2	55	50.013401	92	105.461264	30.193597
3	70	45.696322	92	113.461264	30.632114
4	53	50.504232	92	104.461264	29.889149
...	...	...	...	...	...
76	322	36.900000	50	169.598513	16.132947
77	238	19.197888	115	150.576579	37.923113
78	263	34.000000	50	151.598513	15.769625
79	295	19.833733	119	167.944460	39.423099
80	236	12.101263	107	139.840817	34.948615

81 rows × 5 columns

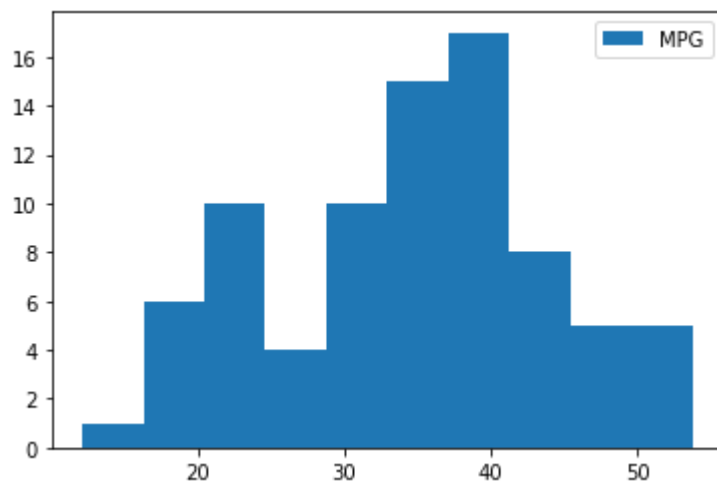
## plotting Distribution for car-MPG

```
In [112]: sns.distplot(Cars.MPG, label='Cars-MPG')  
plt.xlabel('MPG')  
plt.ylabel('Density')  
plt.legend();
```



## Histogram

```
In [113]: plt.hist(Cars.MPG, label='MPG')  
plt.legend();
```



## mean

```
In [115]: Cars.MPG.mean()
```

```
Out[115]: 34.422075728024666
```

## Median

```
In [116]: Cars.MPG.median()
```

```
Out[116]: 35.15272697
```

## Q 21(b)

### import Dataset

```
In [117]: wcat = pd.read_csv('wc-at.csv')  
wcat
```

Out[117]:

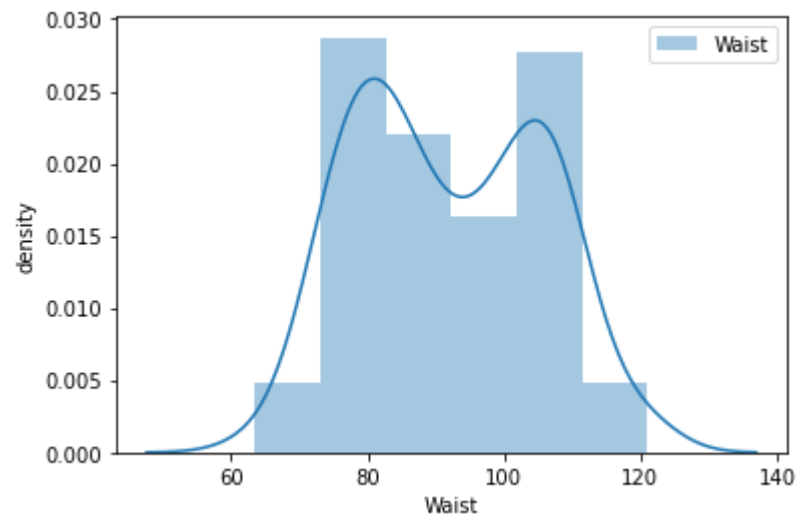
	Waist	AT
0	74.75	25.72
1	72.60	25.89
2	81.80	42.60
3	83.95	42.80
4	74.65	29.84
...	...	...
104	100.10	124.00
105	93.30	62.20
106	101.80	133.00
107	107.90	208.00
108	108.50	208.00

109 rows × 2 columns

### Plotting Distribution for Waist Circumference (Waist)



```
In [118]: sns.distplot(wcat.Waist, label='Waist')  
plt.ylabel('density')  
plt.legend();
```



## Mean

```
In [119]: wcat.Waist.mean()
```

```
Out[119]: 91.90183486238533
```

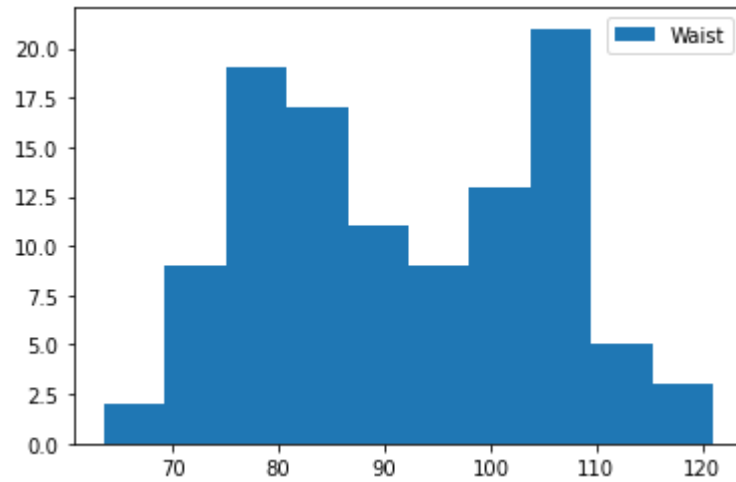
## Median

```
In [120]: wcat.Waist.median()
```

```
Out[120]: 90.8
```

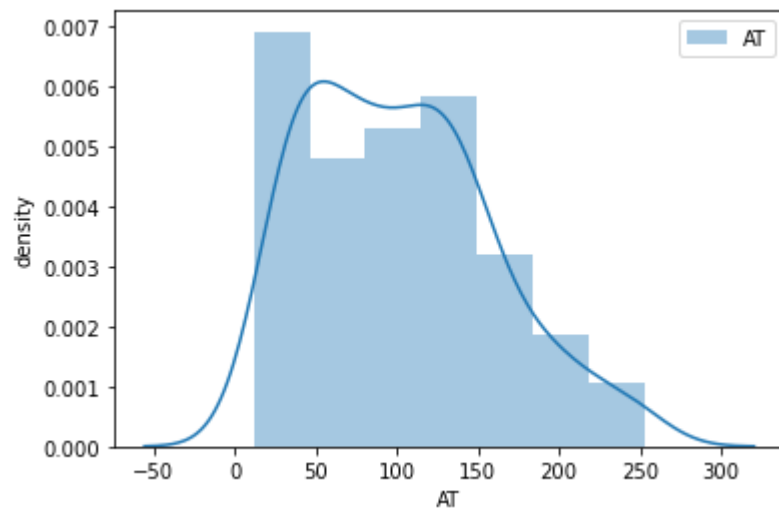
## Histogram

```
In [121]: plt.hist(wcat.Waist,label='Waist')  
plt.legend();
```



## Plotting Distribution for Adipose Tissue (AT)

```
In [122]: sns.distplot(wcat.AT, label='AT')  
plt.ylabel('density')  
plt.legend();
```



## Mean

```
In [123]: wcat.AT.mean()
```

```
Out[123]: 101.89403669724771
```

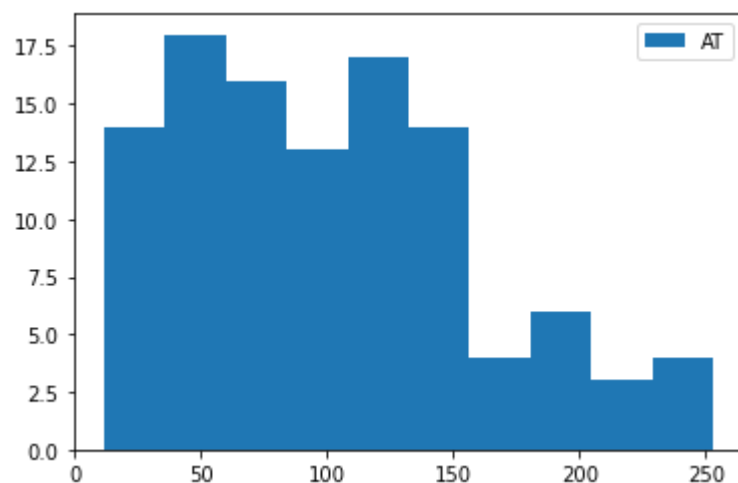
## Median

```
In [125]: wcat.AT.median()
```

```
Out[125]: 96.54
```

## Histogram

```
In [126]: plt.hist(wcat.AT,label='AT')  
plt.legend();
```



=====

**Q\_22.Calculate the Z scores of 90% confidence interval,94% confidence interval, 60% confidence interval**

```
In [127]: # Area Left(AL) = (1+CI) / 2
```

```
In [128]: a = (1+.90)/2  
a
```

```
Out[128]: 0.95
```

```
In [129]: b = (1+.94)/2  
b
```

```
Out[129]: 0.97
```

```
In [130]: c = (1+.6)/2  
c
```

```
Out[130]: 0.8
```

## Z-scores of 90% confidence interval

```
In [131]: stats.norm.ppf(.97)
```

```
Out[131]: 1.8807936081512509
```

## Z-scores of 60% confidence interval

```
In [132]: stats.norm.ppf(0.8)
```

```
Out[132]: 0.8416212335729143
```

=====



**Q\_23. Calculate the Z scores of 90% confidence interval,94% confidence interval, 60% confidence interval**

**t scores of 95% confidence interval for sample size of 25**

```
In [133]: stats.t.ppf(.975,24)
```

```
Out[133]: 2.0638985616280205
```

## t scores of 96% confidence interval for sample size of 2

```
In [134]: stats.t.ppf(.98,24)
```

```
Out[134]: 2.1715446760080677
```

## t scores of 99% confidence interval for sample size of 2

```
In [135]: stats.t.ppf(.995,24)
```

```
Out[135]: 2.796939504772804
```



**Q\_24.A Government company claims that an average light bulb lasts 270 days. A researcher randomly selects 18 bulbs for testing. The sampled bulbs last an average of 260 days, with a standard deviation of 90 days. If the CEO's claim were true, what is the probability that 18 randomly selected bulbs would have an average life of no more than 260 days**

Hint: rcode    `pt(tscore,df)`    df    degrees of freedom

**Assume Null Hypothesis is:  $H_0 = \text{Avg life of Bulb} \geq 260$  days**

**Alternate Hypothesis is:  $H_1 = \text{Avg life of Bulb} < 260 \text{ days}$**

**find t-scores at  $x=260$ ;  $t=(s\_mean-P\_mean)/(s\_SD/\sqrt{n})$**

```
In [136]: t=(260-270)/(90/18**0.5)
          t
```

```
Out[136]: -0.4714045207910317
```

**Find  $P(X \geq 260)$  for null hypothesis**

**$P\_Value = 1 - \text{stats.t.cdf}(\text{abs}(t\_scores), df = n-1)$ ... Using cdf function**

```
In [137]: P_Value = 1-stats.t.cdf(abs(-.4714),df=17)
          P_Value
```

```
Out[137]: 0.32167411684460556
```

**OR  $P\_Value = \text{stats.t.sf}(\text{abs}(t\_score), df = n-1)$ ... Using sf function**

```
In [138]: P_Value=stats.t.sf(abs(-0.4714),df=17)
          P_Value
```

```
Out[138]: 0.32167411684460556
```

=====

◀ ◻ ▶

```
In [ ]:
```

