

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

```
df=pd.read_csv(r"C:\Users\Arigala.Adarsh\Downloads\mtcars.csv")
df
```

| | model | mpg | cyl | disp | hp | drat | wt | qsec | vs |
|----|---------------------|------|-----|-------|-----|------|-------|-------|----|
| 0 | Mazda RX4 | 21.0 | 6 | 160.0 | 110 | 3.90 | 2.620 | 16.46 | 0 |
| 1 | Mazda RX4 Wag | 21.0 | 6 | 160.0 | 110 | 3.90 | 2.875 | 17.02 | 0 |
| 2 | Datsun 710 | 22.8 | 4 | 108.0 | 93 | 3.85 | 2.320 | 18.61 | 1 |
| 3 | Hornet 4 Drive | 21.4 | 6 | 258.0 | 110 | 3.08 | 3.215 | 19.44 | 1 |
| 4 | Hornet Sportabout | 18.7 | 8 | 360.0 | 175 | 3.15 | 3.440 | 17.02 | 0 |
| 5 | Valiant | 18.1 | 6 | 225.0 | 105 | 2.76 | 3.460 | 20.22 | 1 |
| 6 | Duster 360 | 14.3 | 8 | 360.0 | 245 | 3.21 | 3.570 | 15.84 | 0 |
| 7 | Merc 240D | 24.4 | 4 | 146.7 | 62 | 3.69 | 3.190 | 20.00 | 1 |
| 8 | Merc 230 | 22.8 | 4 | 140.8 | 95 | 3.92 | 3.150 | 22.90 | 1 |
| 9 | Merc 280 | 19.2 | 6 | 167.6 | 123 | 3.92 | 3.440 | 18.30 | 1 |
| 10 | Merc 280C | 17.8 | 6 | 167.6 | 123 | 3.92 | 3.440 | 18.90 | 1 |
| 11 | Merc 450SE | 16.4 | 8 | 275.8 | 180 | 3.07 | 4.070 | 17.40 | 0 |
| 12 | Merc 450SL | 17.3 | 8 | 275.8 | 180 | 3.07 | 3.730 | 17.60 | 0 |
| 13 | Merc 450SLC | 15.2 | 8 | 275.8 | 180 | 3.07 | 3.780 | 18.00 | 0 |
| 14 | Cadillac Fleetwood | 10.4 | 8 | 472.0 | 205 | 2.93 | 5.250 | 17.98 | 0 |
| 15 | Lincoln Continental | 10.4 | 8 | 460.0 | 215 | 3.00 | 5.424 | 17.82 | 0 |
| 16 | Chrysler Imperial | 14.7 | 8 | 440.0 | 230 | 3.23 | 5.345 | 17.42 | 0 |
| 17 | Fiat 128 | 32.4 | 4 | 78.7 | 66 | 4.08 | 2.200 | 19.47 | 1 |
| 18 | Honda Civic | 30.4 | 4 | 75.7 | 52 | 4.93 | 1.615 | 18.52 | 1 |
| 19 | Toyota Corolla | 33.9 | 4 | 71.1 | 65 | 4.22 | 1.835 | 19.90 | 1 |

| | | |
|----|---|---|
| 23 | 3 | 4 |
| 24 | 3 | 2 |
| 25 | 4 | 1 |
| 26 | 5 | 2 |
| 27 | 5 | 2 |
| 28 | 5 | 4 |
| 29 | 5 | 6 |
| 30 | 5 | 8 |
| 31 | 4 | 2 |

```
df.shape
```

```
(32, 12)
```

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
RangeIndex: 32 entries, 0 to 31
```

```
Data columns (total 12 columns):
```

| # | Column | Non-Null Count | Dtype |
|----|--------|----------------|---------|
| 0 | model | 32 non-null | object |
| 1 | mpg | 32 non-null | float64 |
| 2 | cyl | 32 non-null | int64 |
| 3 | disp | 32 non-null | float64 |
| 4 | hp | 32 non-null | int64 |
| 5 | drat | 32 non-null | float64 |
| 6 | wt | 32 non-null | float64 |
| 7 | qsec | 32 non-null | float64 |
| 8 | vs | 32 non-null | int64 |
| 9 | am | 32 non-null | int64 |
| 10 | gear | 32 non-null | int64 |
| 11 | carb | 32 non-null | int64 |

```
dtypes: float64(5), int64(6), object(1)
```

```
memory usage: 3.1+ KB
```

```
df.describe()
```

| | mpg | cyl | disp | hp | drat |
|-------|-----------|-----------|------------|------------|-----------|
| count | 32.000000 | 32.000000 | 32.000000 | 32.000000 | 32.000000 |
| mean | 20.090625 | 6.187500 | 230.721875 | 146.687500 | 3.596563 |
| std | 6.026948 | 1.785922 | 123.938694 | 68.562868 | 0.534679 |
| min | 10.400000 | 4.000000 | 71.100000 | 52.000000 | 2.760000 |
| 25% | 15.425000 | 4.000000 | 120.825000 | 96.500000 | 3.080000 |
| 50% | 19.200000 | 6.000000 | 196.300000 | 123.000000 | 3.695000 |

```

3.325000
75%    22.800000    8.000000    326.000000    180.000000    3.920000
3.610000
max     33.900000    8.000000    472.000000    335.000000    4.930000
5.424000

```

| | qsec | vs | am | gear | carb |
|-------|-----------|-----------|-----------|-----------|---------|
| count | 32.000000 | 32.000000 | 32.000000 | 32.000000 | 32.0000 |
| mean | 17.848750 | 0.437500 | 0.406250 | 3.687500 | 2.8125 |
| std | 1.786943 | 0.504016 | 0.498991 | 0.737804 | 1.6152 |
| min | 14.500000 | 0.000000 | 0.000000 | 3.000000 | 1.0000 |
| 25% | 16.892500 | 0.000000 | 0.000000 | 3.000000 | 2.0000 |
| 50% | 17.710000 | 0.000000 | 0.000000 | 4.000000 | 2.0000 |
| 75% | 18.900000 | 1.000000 | 1.000000 | 4.000000 | 4.0000 |
| max | 22.900000 | 1.000000 | 1.000000 | 5.000000 | 8.0000 |

```
df.isnull().sum()
```

```

model      0
mpg        0
cyl        0
disp       0
hp         0
drat       0
wt         0
qsec       0
vs         0
am         0
gear       0
carb       0
dtype: int64

```

```
df.duplicated().sum()
```

```
0
```

```
df.mean()
```

```

C:\Users\Arigala.Adarsh\AppData\Local\Temp\
ipykernel_24668\3698961737.py:1: FutureWarning: Dropping of nuisance
columns in DataFrame reductions (with 'numeric_only=None') is
deprecated; in a future version this will raise TypeError.  Select
only valid columns before calling the reduction.

```

```
df.mean()
```

```

mpg      20.090625
cyl       6.187500
disp     230.721875
hp       146.687500
drat      3.596563
wt        3.217250

```

```
qsec      17.848750
vs         0.437500
am         0.406250
gear       3.687500
carb       2.812500
dtype: float64
```

Mean

```
df.mpg.mean()
```

20.090624999999996

Median

```
df.mpg.median()
```

19.2

Mode

```
df.mpg.mode()
```

| | |
|---|------|
| 0 | 10.4 |
| 1 | 15.2 |
| 2 | 19.2 |
| 3 | 21.0 |
| 4 | 21.4 |
| 5 | 22.8 |
| 6 | 30.4 |

```
Name: mpg, dtype: float64
```

```
df[df.mpg==10.4]
```

| | model | mpg | cyl | disp | hp | drat | wt | qsec | vs |
|---------|---------------------|------|-----|-------|-----|------|-------|-------|----|
| am \ | Cadillac Fleetwood | 10.4 | 8 | 472.0 | 205 | 2.93 | 5.250 | 17.98 | 0 |
| 14 0 | | | | | | | | | |
| 15 0 | Lincoln Continental | 10.4 | 8 | 460.0 | 215 | 3.00 | 5.424 | 17.82 | 0 |

| | gear | carb |
|----|------|------|
| 14 | 3 | 4 |
| 15 | 3 | 4 |

```
df[df.mpg==30.4]
```

[illegible]

| | | | | | | | | | | | |
|----|-------|--------|------|---|------|-----|------|-------|-------|---|---|
| 27 | Lotus | Europa | 30.4 | 4 | 95.1 | 113 | 3.77 | 1.513 | 16.90 | 1 | 1 |
| 5 | 2 | | | | | | | | | | |

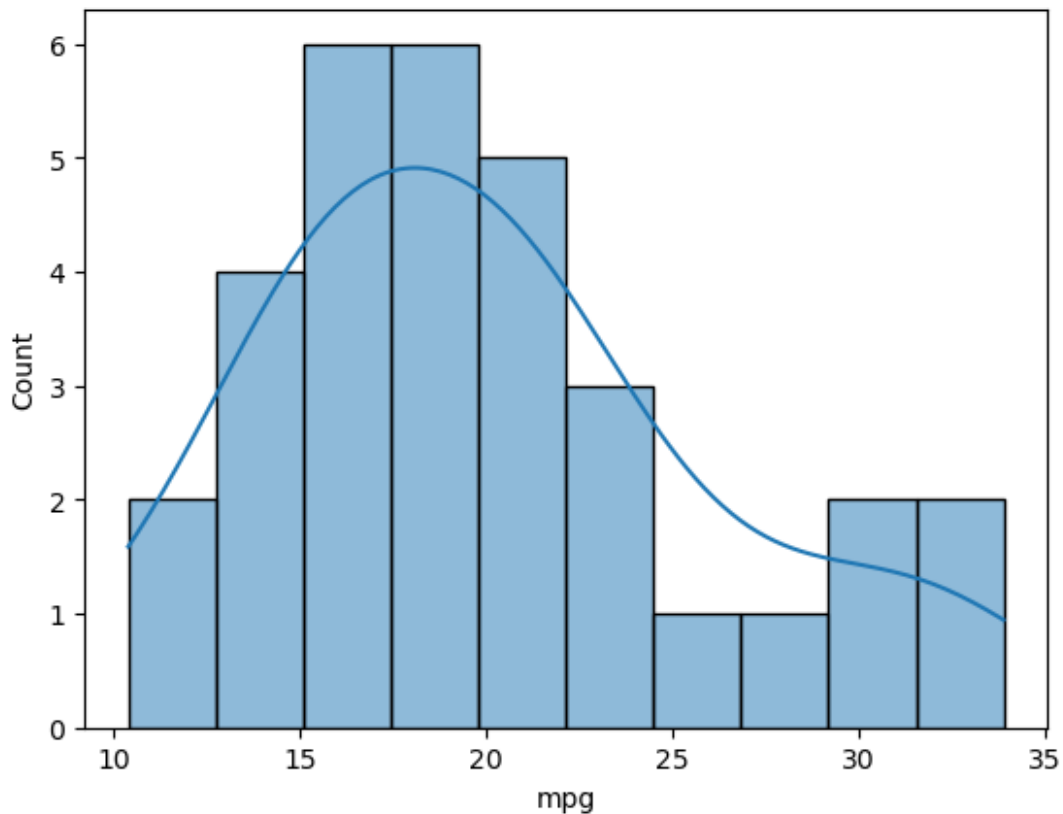
Range

```
max=df.mpg.max()
min=df.mpg.min()

range=max-min
print("Range of mpg:",range)
```

Range of mpg: 23.5

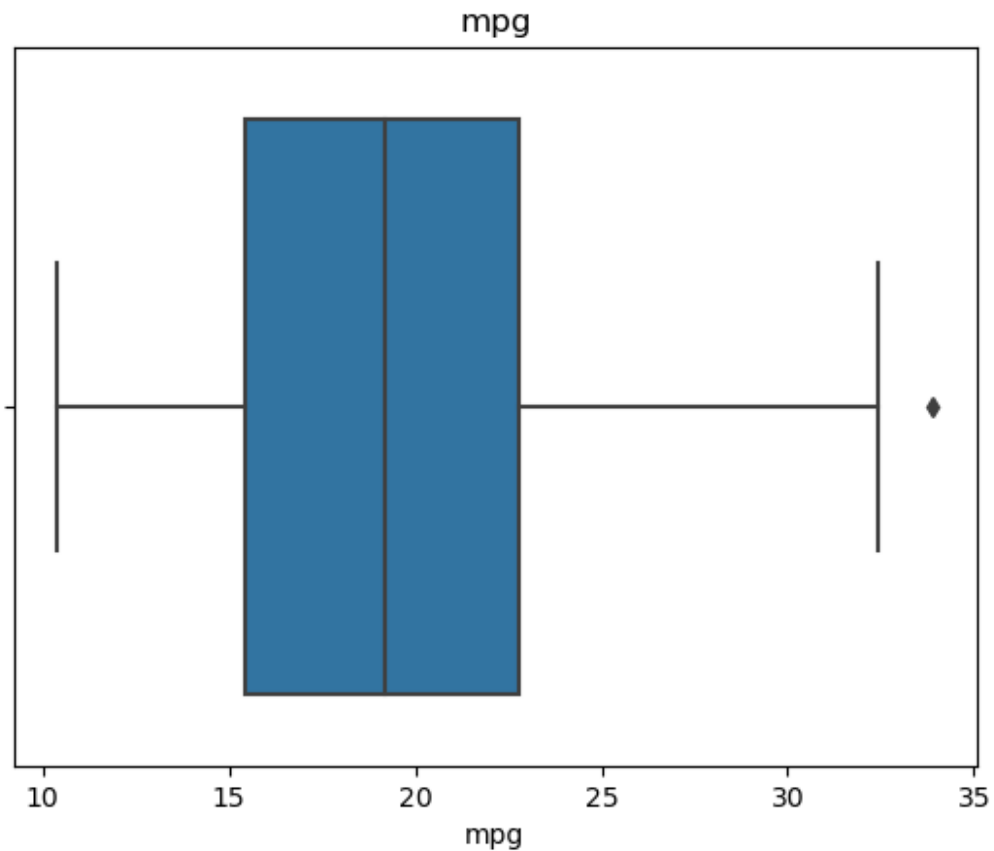
```
sns.histplot(df.mpg,bins=10,kde=True)
plt.show()
```



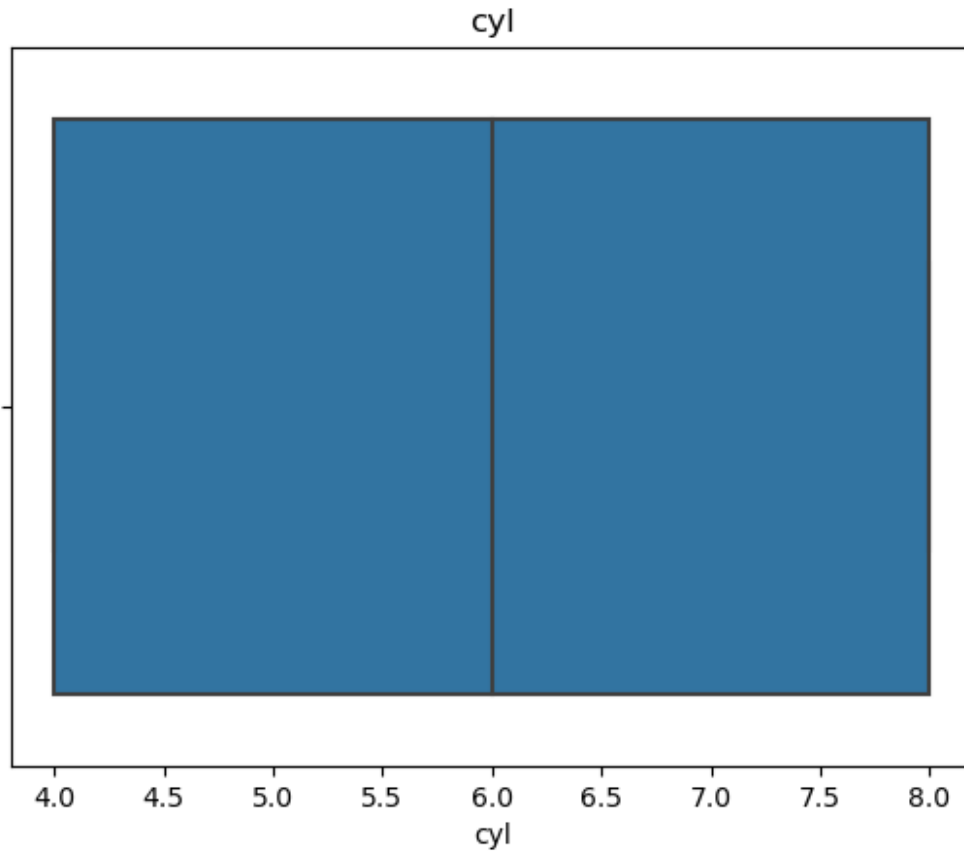
```
for i in df.columns:
    if(df[i].dtypes!='object'):
        sns.boxplot(df[i])
        plt.title(i)
        plt.show()
```

C:\Users\Arigala.Adarsh\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument

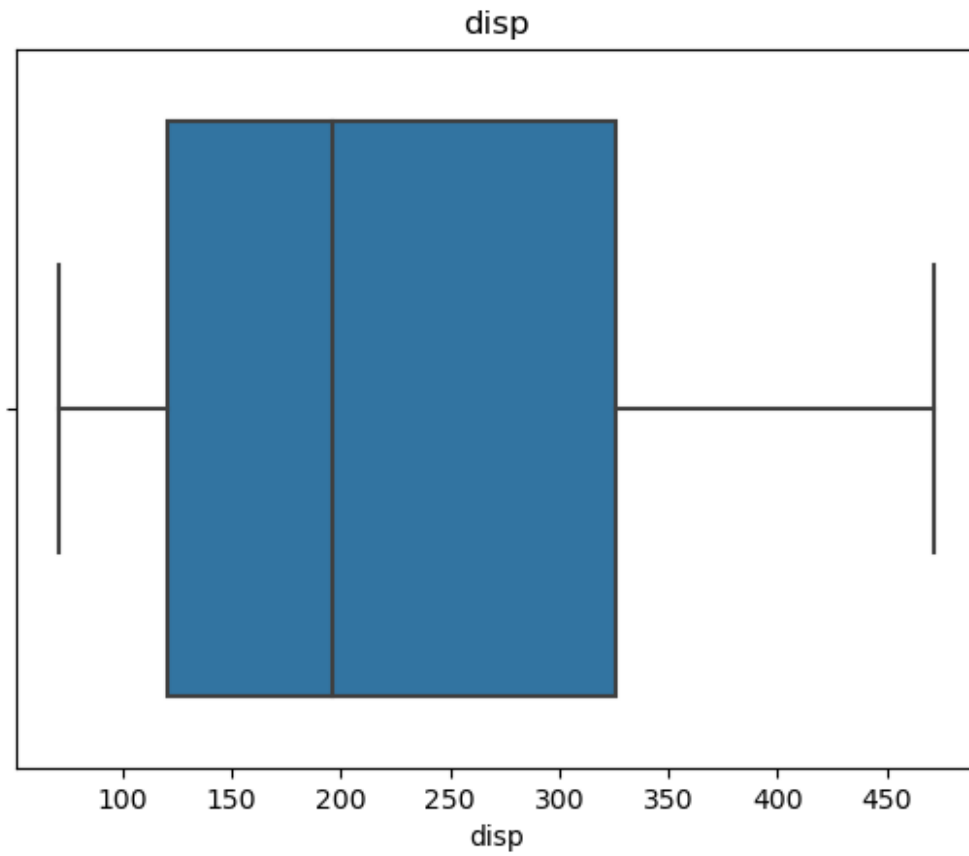
will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.
warnings.warn()



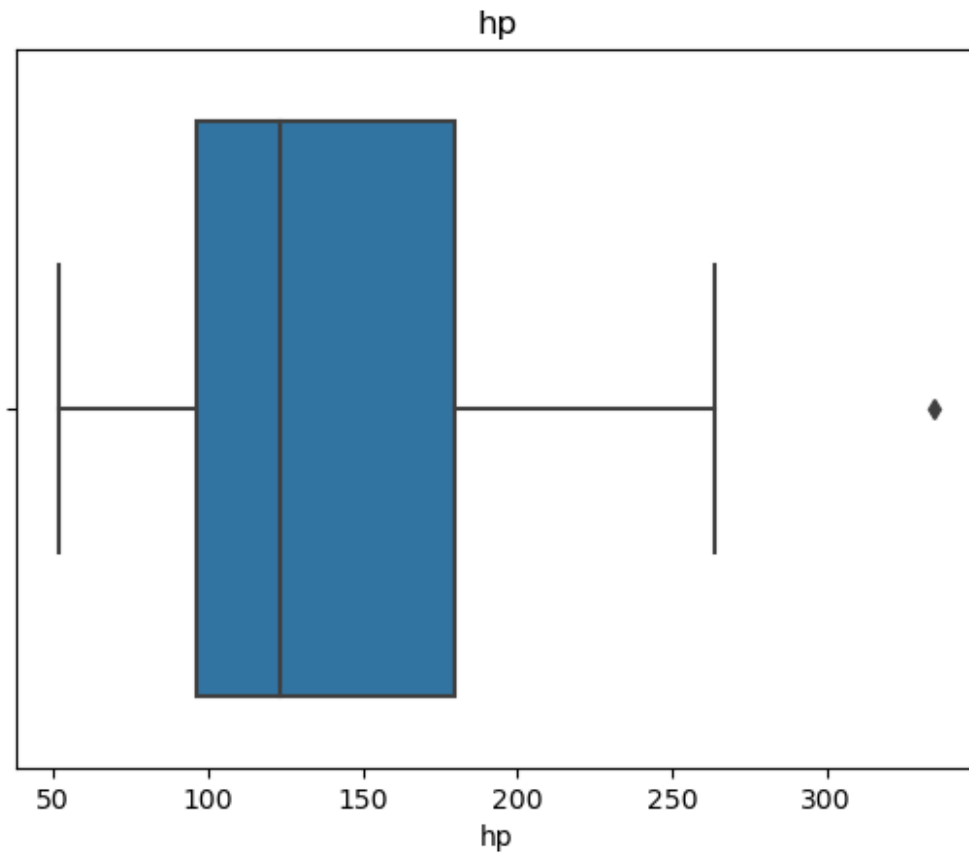
C:\Users\Arigala.Adarsh\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.
warnings.warn()



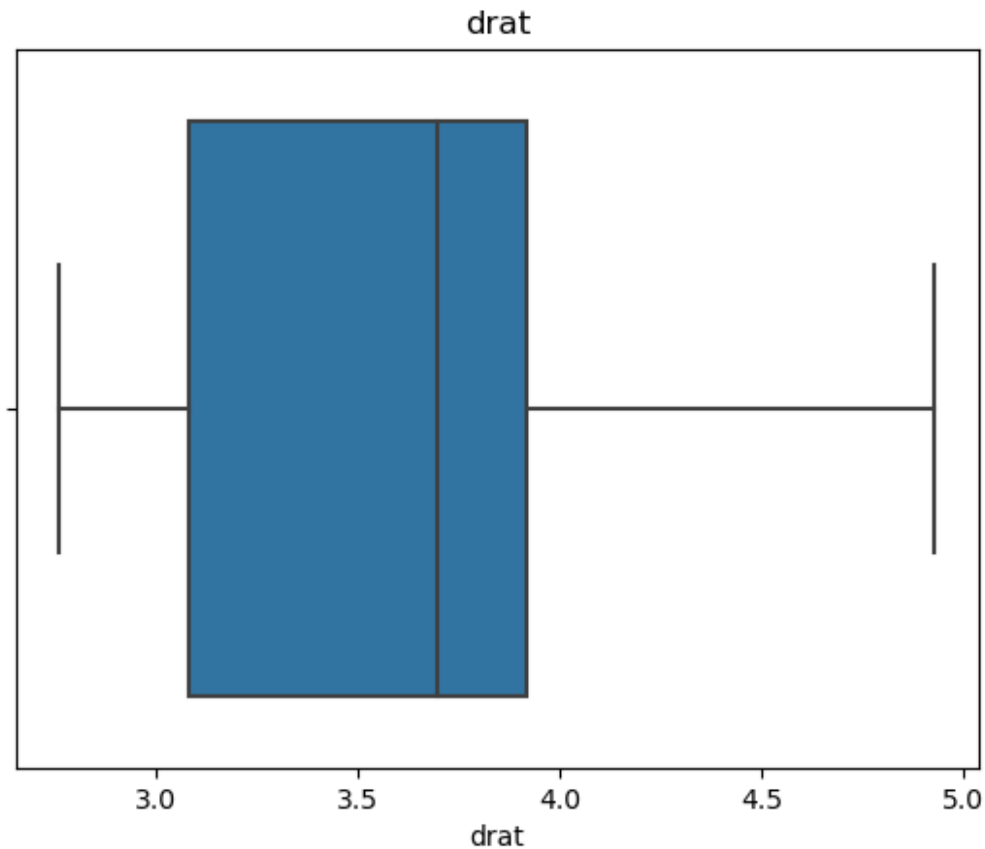
```
C:\Users\Arigala.Adarsh\anaconda3\lib\site-packages\seaborn\
_decorators.py:36: FutureWarning: Pass the following variable as a
keyword arg: x. From version 0.12, the only valid positional argument
will be `data`, and passing other arguments without an explicit
keyword will result in an error or misinterpretation.
  warnings.warn(
```

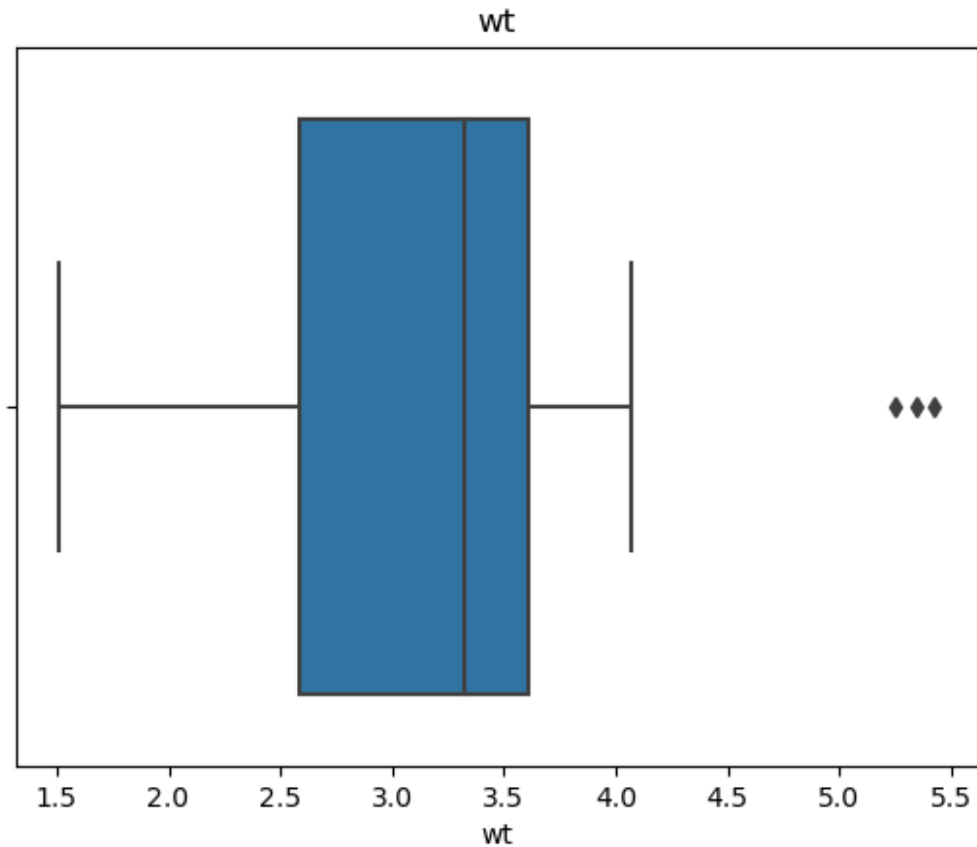
```
C:\Users\Arigala.Adarsh\anaconda3\lib\site-packages\seaborn\
_decorators.py:36: FutureWarning: Pass the following variable as a
keyword arg: x. From version 0.12, the only valid positional argument
will be `data`, and passing other arguments without an explicit
keyword will result in an error or misinterpretation.
  warnings.warn(
```



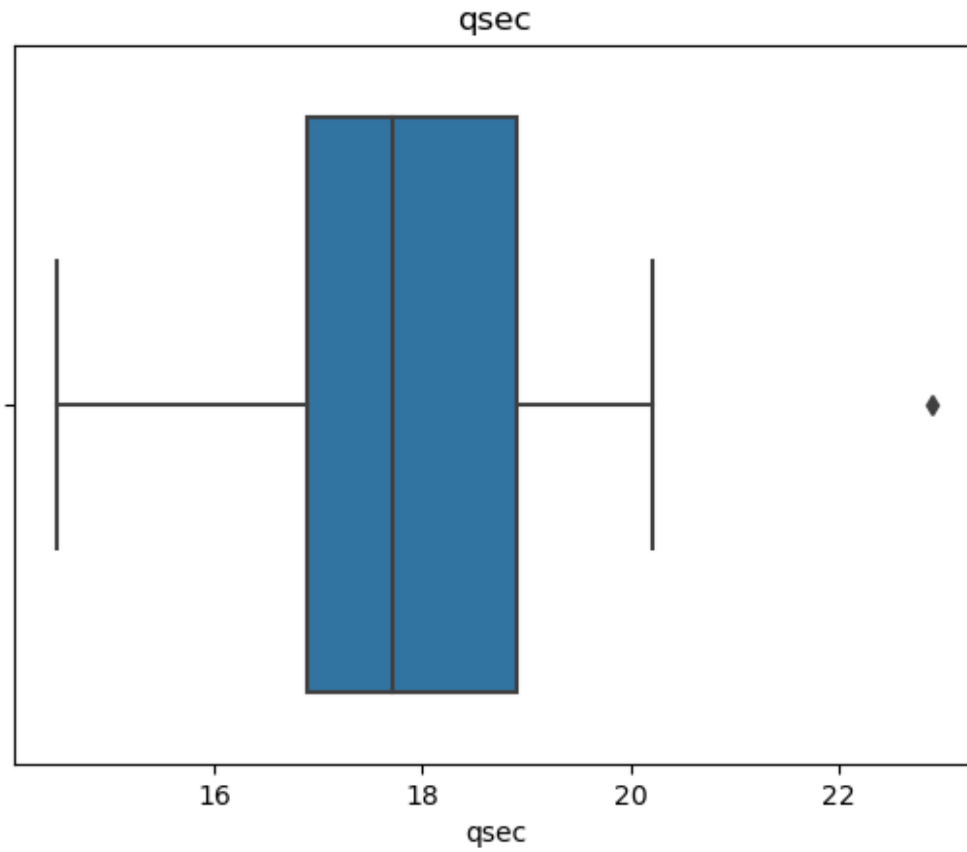
```
C:\Users\Arigala.Adarsh\anaconda3\lib\site-packages\seaborn\
_decorators.py:36: FutureWarning: Pass the following variable as a
keyword arg: x. From version 0.12, the only valid positional argument
will be `data`, and passing other arguments without an explicit
keyword will result in an error or misinterpretation.
  warnings.warn(
```



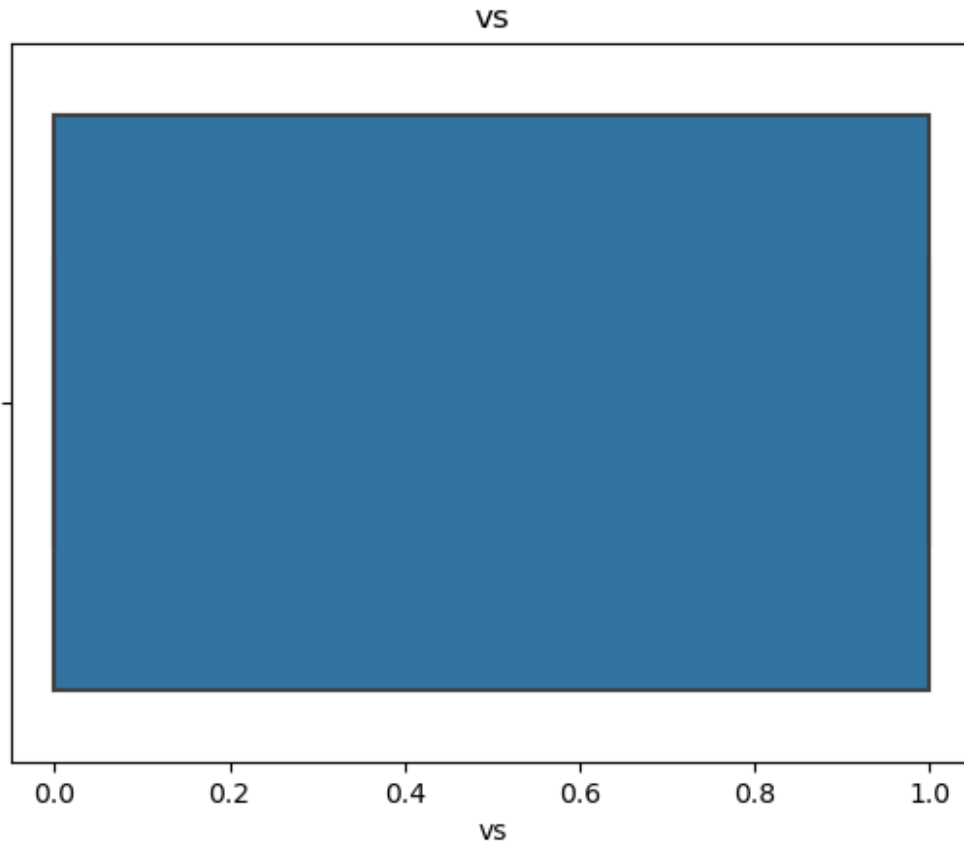
```
C:\Users\Arigala.Adarsh\anaconda3\lib\site-packages\seaborn\
_decorators.py:36: FutureWarning: Pass the following variable as a
keyword arg: x. From version 0.12, the only valid positional argument
will be `data`, and passing other arguments without an explicit
keyword will result in an error or misinterpretation.
  warnings.warn(
```



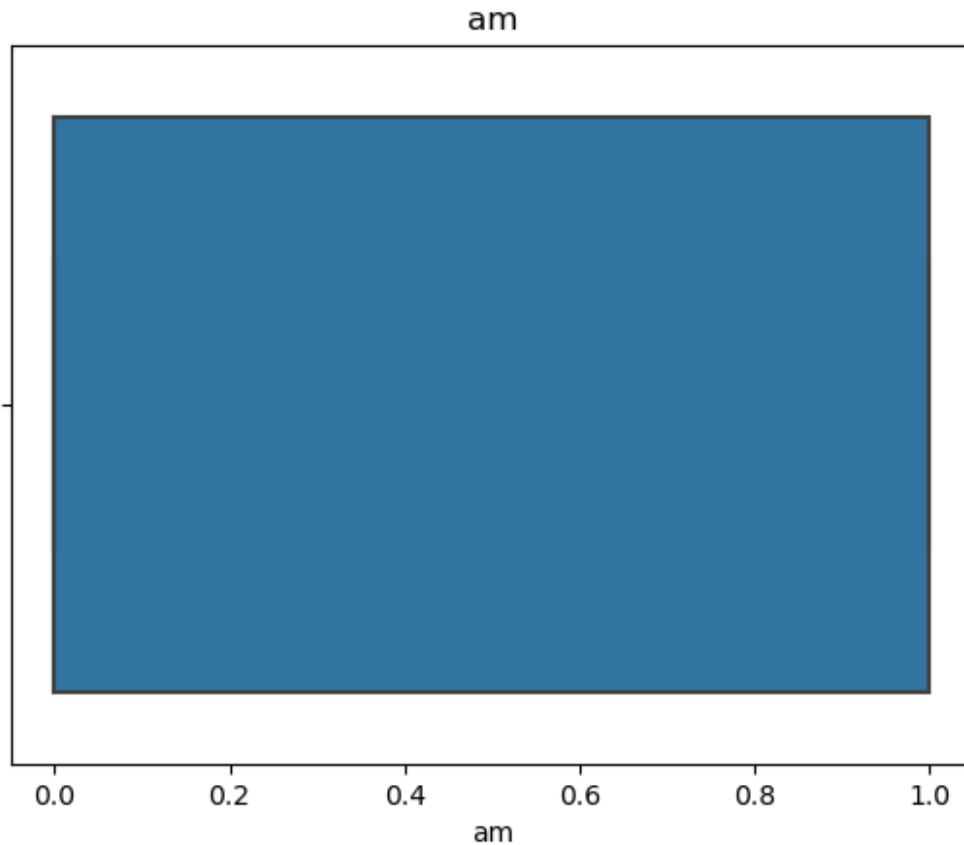
```
C:\Users\Arigala.Adarsh\anaconda3\lib\site-packages\seaborn\
_decorators.py:36: FutureWarning: Pass the following variable as a
keyword arg: x. From version 0.12, the only valid positional argument
will be `data`, and passing other arguments without an explicit
keyword will result in an error or misinterpretation.
  warnings.warn(
```



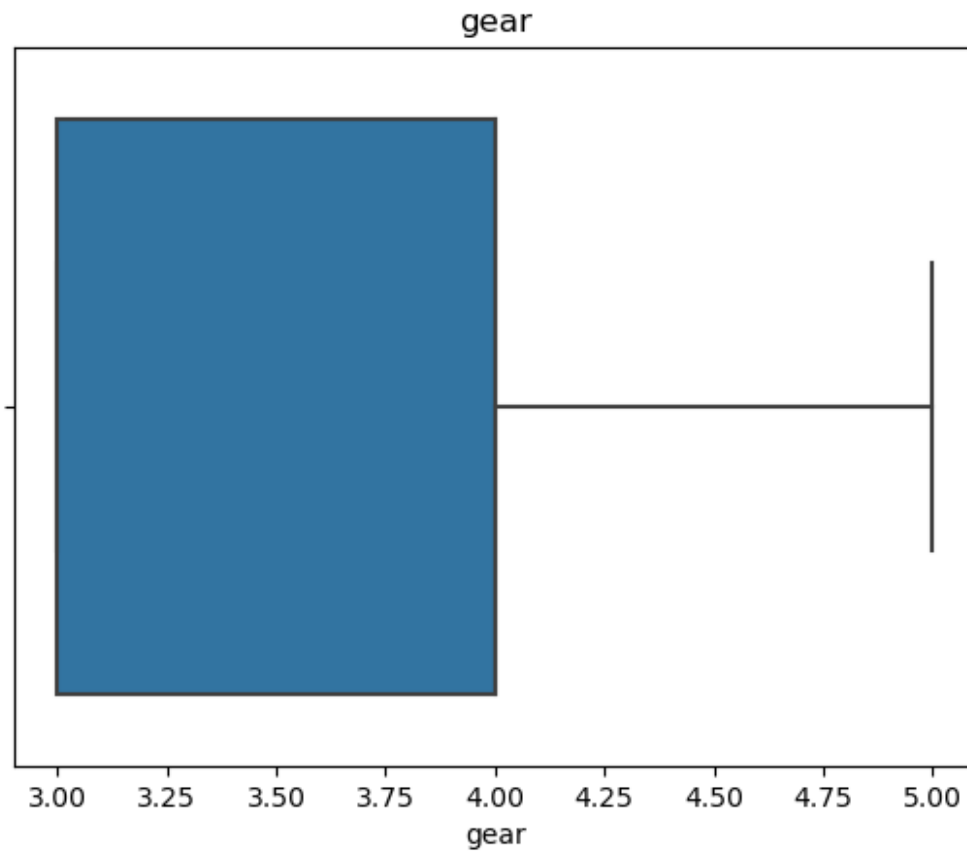
```
C:\Users\Arigala.Adarsh\anaconda3\lib\site-packages\seaborn\
_decorators.py:36: FutureWarning: Pass the following variable as a
keyword arg: x. From version 0.12, the only valid positional argument
will be `data`, and passing other arguments without an explicit
keyword will result in an error or misinterpretation.
  warnings.warn(
```



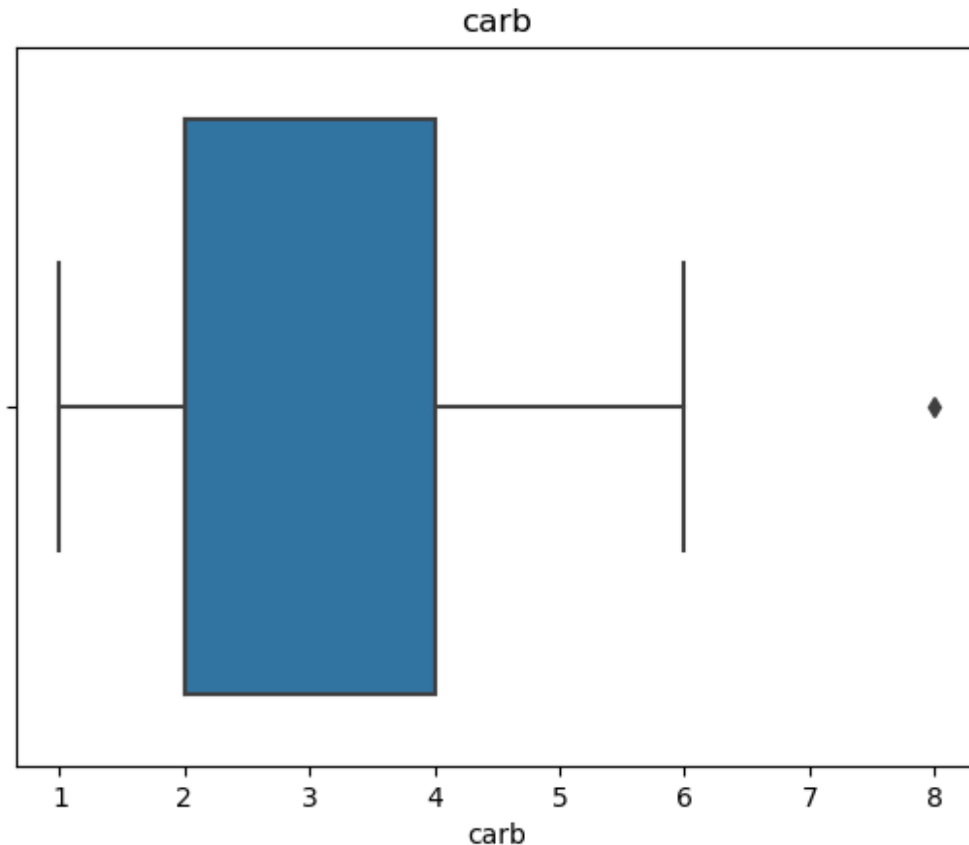
```
C:\Users\Arigala.Adarsh\anaconda3\lib\site-packages\seaborn\
_decorators.py:36: FutureWarning: Pass the following variable as a
keyword arg: x. From version 0.12, the only valid positional argument
will be `data`, and passing other arguments without an explicit
keyword will result in an error or misinterpretation.
  warnings.warn(
```



```
C:\Users\Arigala.Adarsh\anaconda3\lib\site-packages\seaborn\
_decorators.py:36: FutureWarning: Pass the following variable as a
keyword arg: x. From version 0.12, the only valid positional argument
will be `data`, and passing other arguments without an explicit
keyword will result in an error or misinterpretation.
  warnings.warn(
```



```
C:\Users\Arigala.Adarsh\anaconda3\lib\site-packages\seaborn\
_decorators.py:36: FutureWarning: Pass the following variable as a
keyword arg: x. From version 0.12, the only valid positional argument
will be `data`, and passing other arguments without an explicit
keyword will result in an error or misinterpretation.
  warnings.warn(
```

Variance and Standard Deviation

```
print("Variance of mpg is :",df.mpg.var())  
print("Standard deviation of mpg is :",df.mpg.std())
```

```
Variance of mpg is : 36.32410282258065  
Standard deviation of mpg is : 6.026948052089105
```

Inferential Statistics

```
from scipy import stats
```

```
#pip install bioinfokit
```

```
from bioinfokit.analys import get_data
```

One Sample Z-test

One Sample Z-test checks whether the sample comes from a known population where population mean and standard deviation (σ) should be known.

```
df1=get_data('z_one_samp').data  
df1
```

| | sizes |
|----|----------|
| 0 | 4.819289 |
| 1 | 3.569358 |
| 2 | 5.346402 |
| 3 | 5.950908 |
| 4 | 5.871183 |
| 5 | 5.590780 |
| 6 | 3.591993 |
| 7 | 5.137837 |
| 8 | 3.870271 |
| 9 | 4.624155 |
| 10 | 5.796371 |
| 11 | 6.194647 |
| 12 | 4.167342 |
| 13 | 5.168888 |
| 14 | 6.646000 |
| 15 | 4.327118 |
| 16 | 4.609097 |
| 17 | 6.828167 |
| 18 | 3.407583 |
| 19 | 4.218922 |
| 20 | 4.413460 |
| 21 | 4.034847 |
| 22 | 5.979468 |
| 23 | 4.470525 |
| 24 | 4.146668 |
| 25 | 4.711762 |
| 26 | 4.918706 |
| 27 | 5.297779 |
| 28 | 5.473580 |
| 29 | 5.813743 |
| 30 | 4.189742 |
| 31 | 4.522986 |
| 32 | 6.333137 |
| 33 | 5.961645 |
| 34 | 5.606989 |
| 35 | 4.559243 |
| 36 | 4.135710 |
| 37 | 6.597137 |

```

38 4.261241
39 2.776458
40 6.507718
41 4.905230
42 4.662474
43 4.829080
44 6.308754
45 5.198246
46 5.222804
47 6.362502
48 2.906932
49 6.053046

```

- Factory produces balls of a diameter of 5 cm
- Std. deviation = **0.4**
- Quality officer wants to test whether the ball diameter is significantly different from 5 cm or not
- Null Hypothesis - Sample Mean = 5 cm |||| Alternate hypothesis - Sample Mean not equal to 5

```

from bioinfokit.analys import stat
res=stat()
res.ztest(df=df1,x="sizes",mu=5,x_std=0.4,test_type=1)
print(res.summary)

```

One Sample Z-test

```

-----
Sample size      50
Mean            5.01796
Z value         0.317465
p value (one-tail) 0.375446
p value (two-tail) 0.750891
Lower 95.0%     4.90709
Upper 95.0%     5.12883
-----

```

- we can conclude that z_test is not significant. The mean diameter of the balls in random sample is equal to the population mean is 5cm. Accept the null hypothesis.

Two Sample Z-Test

Null hypothesis: Two group means are equal Alternate hypothesis: Two group means are different (not equal)

```
df2 = get_data('z_two_samp').data
df2.head()
```

| | fact_A | fact_B |
|---|----------|----------|
| 0 | 4.977904 | 5.887947 |
| 1 | 5.166254 | 5.990616 |
| 2 | 4.991749 | 6.110116 |
| 3 | 4.901557 | 5.936784 |
| 4 | 4.713866 | 6.227506 |

Std. deviation of two population is fixed at 0.1

```
res = stat()
res.ztest(df=df2, x='fact_A', y='fact_B', x_std=0.1, y_std=0.1,
test_type=2)
```

```
print(res.summary)
```

Two Sample Z-test

| | |
|--------------------|-----------|
| Sample size for x | 50 |
| Sample size for y | 50 |
| Mean of x | 5.01284 |
| Mean of y | 5.99015 |
| Z value | -48.8656 |
| p value (one-tail) | 0 |
| p value (two-tail) | 0 |
| Lower 95.0% | -1.01651 |
| Upper 95.0% | -0.938113 |

- Reject the null hypothesis and can conclude that there is a significant difference in the ball size produced in factories A and B
- Factory A is closer to mean 5 however Factory B is not

T Test

One Sample Test

- used for comparing the sample mean (a random sample from a population) with the specific value
- In t-test, the population variance is unknown and it is estimated from sample variance
- If $n < 30$, Z-test cant be applied
- Null Hypothesis: Sample mean is equal to the hypothesized or known population mean
- Alternate Hypothesis: Sample mean is not equal to the hypothesized or known population mean

```

t1 = get_data('t_one_samp').data
t1.head(2)

      size
0  5.739987
1  5.254042

res=stat()
res.ttest(df=t1, test_type=1, res='size', mu=5)

print(res.summary)

```

One Sample t-test

```

-----
Sample size      50
Mean             5.05128
t                0.36789
Df              49
p value (one-tail) 0.35727
p value (two-tail) 0.71454
Lower 95.0%      4.77116
Upper 95.0%      5.3314
-----

```

Two Sample T-Test

t-test compares the means of two independent groups, determining whether they are equal or significantly different

```

### load dataset as pandas dataframe

t2 = get_data('t_ind_samp').data
t2

```

| | Genotype | yield |
|----|----------|-------|
| 0 | A | 78.0 |
| 1 | A | 84.3 |
| 2 | A | 81.0 |
| 3 | B | 88.0 |
| 4 | B | 92.0 |
| 5 | B | 84.1 |
| 6 | A | 74.5 |
| 7 | A | 77.8 |
| 8 | A | 79.0 |
| 9 | B | 88.0 |
| 10 | B | 92.5 |
| 11 | B | 91.8 |

```
res = stat()
res.ttest(df=t2, xfac='Genotype', res='yield', test_type=2)

print(res.summary)
```

Two sample t-test with equal variance

```
-----
Mean diff          -10.3
t                  -5.40709
Std Error           1.90491
df                  10
p value (one-tail)  0.000149204
p value (two-tail)  0.000298408
Lower 95.0%        -14.5444
Upper 95.0%        -6.05561
-----
```

Parameter estimates

| Level | Number | Mean | Std Dev | Std Error | Lower 95.0% Upper 95.0% |
|-------|--------|------|---------|-----------|----------------------------|
| A | 6 | 79.1 | 3.30817 | 1.35056 | 75.6283 82.5717 |
| B | 6 | 89.4 | 3.29059 | 1.34338 | 85.9467 92.8533 |

- statistically significant and hence reject the null hypothesis

Chi-Square Test

- Test for independence
- Null Hypothesis: there is no association between the two categorical variables
- Alternate Hypothesis: there is an association between the two categorical variables

```
chi = get_data('drugdata').data
chi
```

| | treatments | cured | noncured |
|---|------------|-------|----------|
| 0 | treated | 60 | 10 |
| 1 | nontreated | 30 | 25 |

```
### Set treatment column as your index
chi= chi.set_index('treatments')
```

```
chi
```

```
      cured  noncured
treatments
treated      60      10
nontreated   30      25
```

```
res = stat()
res.chisq(df=chi)
print(res.summary)
```

Chi-squared test for independence

| Test | Df | Chi-square | P-value |
|----------------|------|------------|-------------|
| ----- | ---- | ----- | ----- |
| Pearson | 1 | 13.3365 | 0.000260291 |
| Log-likelihood | 1 | 13.4687 | 0.000242574 |