



Department of Computer Science and Engineering (Data Science)
B.Tech. Sem: III Subject: Statistics for Data Science
Experiment 5

Date:	Experiment Title: Confidence Interval
Aim	To implement Confidence Interval using Python.
Software	Google Colab
Implementation	<pre>import numpy as np import scipy import statistics as st import math import matplotlib.pyplot as plt from scipy.stats import norm</pre> <p>Using Python solve the following questions given below :</p> <p>1. If Z follows standard normal distribution, then find</p> <p>(i) $P(Z < 1.5)$ (ii) $P(Z > 0.5)$ (iii) $P(Z < 1.5)$ (iv) $P(Z > 0.5)$</p> <p>(v) $P(-2.2 < Z < 1)$</p> <p>Code:</p> <pre>P1 = norm.cdf(1.5) print("P(z<1.5) = " + str(P1)) P2 = norm.sf(0.5) print("P(Z> 0.5) = " + str(P2)) P3 = norm.cdf(1.5) - norm.cdf(-1.5) print("P(Z < 1.5) = " + str(P3)) P4 = norm.cdf(-0.5) + norm.sf(0.5) print("P Z > 0.5) = "+ str(P4)) P5 = norm.cdf(1) - norm.cdf(-2.2) print("P(-2.2 < Z < 1) = " + str(P5))</pre> <p>output:</p> <pre>1. P(z<1.5) = 0.9331927987311419 2. P(Z> 0.5) = 0.3085375387259869 3. P(Z < 1.5) = 0.8663855974622838 4. P Z > 0.5) = 0.6170750774519738 5. P(-2.2 < Z < 1) = 0.8274412985550443</pre> <p>2. If Z follows standard normal distribution, then find value of Z_0 satisfying the given equation</p> <p>(i) $P(Z < Z_0) = 0.90$ (ii) $P(Z < Z_0) = 0.95$ (iii) $P(Z < Z_0) = 0.99$</p> <p>(iv) $P(Z < Z_0) = 0.90$ (v) $P(Z < Z_0) = 0.95$ (vi) $P(Z < Z_0) = 0.99$</p> <p>Code:</p> <pre>print("value of z_0 satsifying equation P(z< z_0) = 0.90 is " + str(norm.ppf(0.90))) print("value of z_0 satsifying equation P(z< z_0) = 0.95 is " + str(norm.ppf(0.95)))</pre>

```
print("value of z_0 satsifying equation  $P(z < z_0) = 0.99$  is " + str(norm.ppf(0.99)))
print("value of z_0 satsifying equation  $P(|z| < z_0) = 0.90$  is " + str(norm.ppf(0.95)))
print("value of z_0 satsifying equation  $P(|z| < z_0) = 0.95$  is " + str(norm.ppf(0.975)))
print("value of z_0 satsifying equation  $P(|z| < z_0) = 0.99$  is " + str(norm.ppf(0.995)))

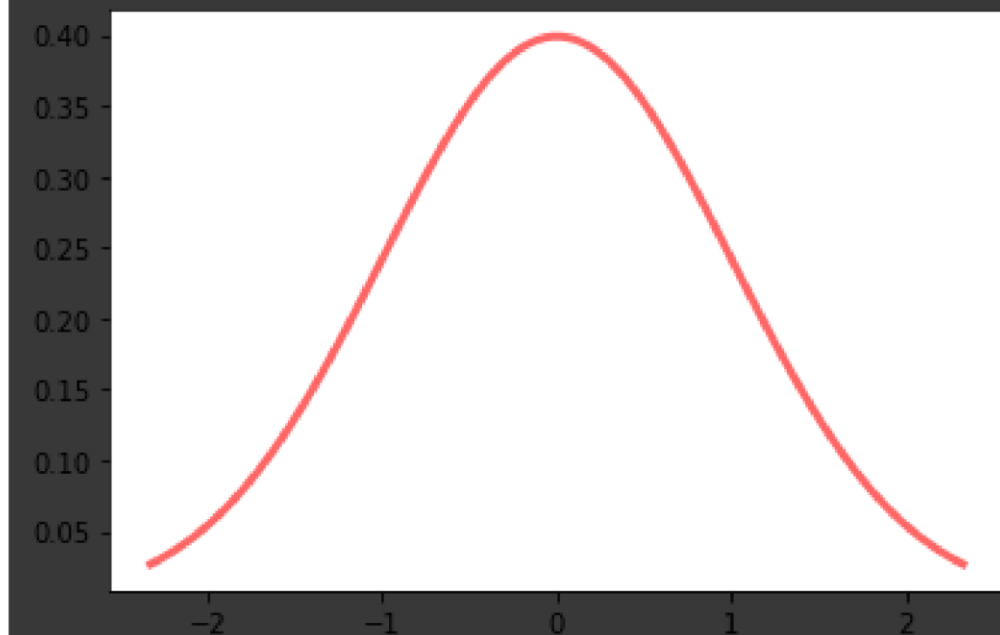
fig, ax = plt.subplots(1,1)
mean, var, skew, kurt = norm.stats(moments = 'mvsk')
print(mean, var, skew, kurt)
x = np.linspace(norm.ppf(0.01), norm.ppf(0.99), 100)
ax.plot(x, norm.pdf(x), 'r', lw=3, alpha = 0.6, label = 'norm pdf')
```

output:

```
value of z_0 satsifying equation  $P(z < z_0) = 0.90$  is 1.2815515655446004
value of z_0 satsifying equation  $P(z < z_0) = 0.95$  is 1.6448536269514722
value of z_0 satsifying equation  $P(z < z_0) = 0.99$  is 2.3263478740408408
value of z_0 satsifying equation  $P(|z| < z_0) = 0.90$  is 1.6448536269514722
value of z_0 satsifying equation  $P(|z| < z_0) = 0.95$  is 1.959963984540054
value of z_0 satsifying equation  $P(|z| < z_0) = 0.99$  is 2.5758293035489004
```

```
0.0 1.0 0.0 0.0
```

```
[<matplotlib.lines.Line2D at 0x7f5cacd7dc50>]
```



3. If t follows students t distribution, then find

(i) $P(t < 1.5)$ with d.o.f. = 20

(ii) $P(t > 0.5)$ with d.o.f. = 15

(iii) $P(|t| < 1.5)$ with d.o.f. = 25

(iv) $P(|t| > 0.5)$ with d.o.f. = 35

(v) $P(-2.2 < t < 1)$ with d.o.f. = 42

Code:

```
from scipy.stats import t
P1 = t.cdf(1.5, 20)
```

```
print("With df = 20, P(t < 1.5) = " + str(P1))
P2 = t.sf(0.5, 15)
print("With df = 15, P(t > 0.5) = " + str(P2))
P3 = t.cdf(1.5, 25) - t.cdf(-1.5, 25)
print("With df = 25, P(t < 1.5) = " + str(P3))
P4 = t.cdf(-0.5, 35) + t.sf(0.5, 35)
print("With df = 35, P(|t| > 0.5) = " + str(P4))
P5 = t.cdf(1, 42) - t.cdf(-2.2, 42)
print("With df = 42, P(-2.2 < t < 1) = " + str(P5))
```

output:

1. With df = 20, P(t < 1.5) = 0.9253821144153737
2. With df = 15, P(t > 0.5) = 0.3121650567600378
3. With df = 25, P(t < 1.5) = 0.8538615348619807
4. With df = 35, P(|t| > 0.5) = 0.6202043032354958
5. With df = 42, P(-2.2 < t < 1) = 0.8218007171327162

4. If t follows students t distribution, then find value of t_0 satisfying the given equation

- (i) $P(t < t_0) = 0.90$ with d.o.f. = 20 (ii) $P(t < t_0) = 0.95$ with d.o.f. = 15
- (iii) $P(t < t_0) = 0.99$ with d.o.f. = 25 (iv) $P(|t| < t_0) = 0.90$ with d.o.f. = 30
- (v) $P(|t| < t_0) = 0.95$ with d.o.f. = 42 (vi) $P(|t| < t_0) = 0.99$ with d.o.f. = 10

Code:

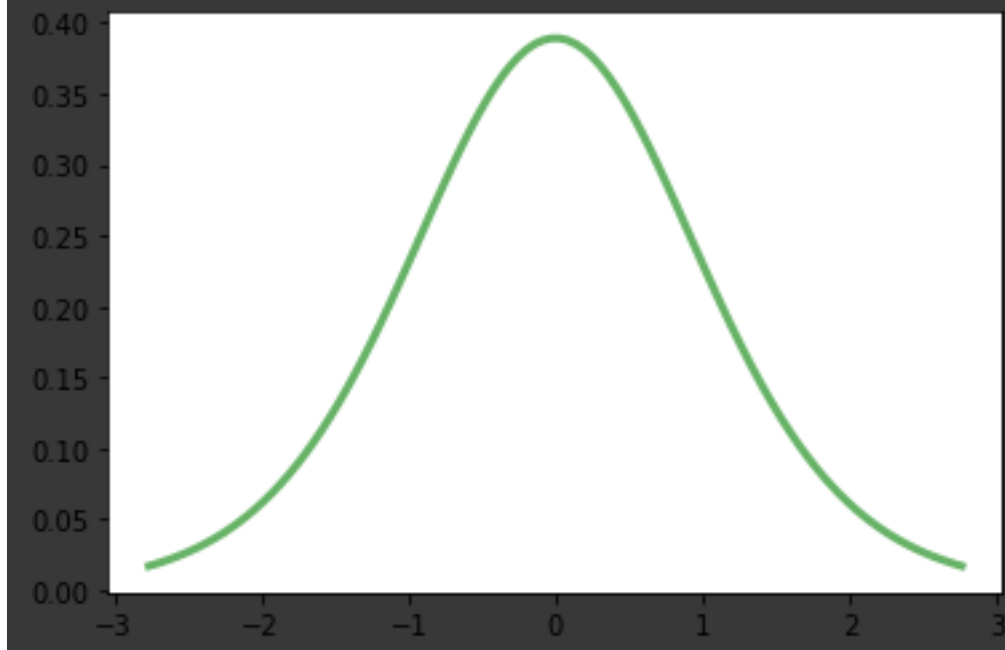
```
print("with df = 20 value of t_0 satisfying equation P(t < t_0) = 90 is " + str(t.ppf(0.90, 20)))
print("with df = 15 value of t_0 satisfying equation P(t < t_0) = 95 is " + str(t.ppf(0.95, 15)))
print("with df = 25 value of t_0 satisfying equation P(t < t_0) = 99 is " + str(t.ppf(0.99, 25)))
print("with df = 30 value of t_0 satisfying equation P(|t| < t_0) = 90 is " + str(t.ppf(0.95, 30)))
print("with df = 42 value of t_0 satisfying equation P(|t| < t_0) = 95 is " + str(t.ppf(0.975, 42)))
```

```
fig,ax = plt.subplots(1,1)
df = 10
mean, var, skew, kurt = t.stats(df, moments='mvsk')
print(mean, var, skew, kurt)
x = np.linspace(t.ppf(0.01, df), t.ppf(0.99, df), 100)
ax.plot(x, t.pdf(x, df), 'g', lw=3, alpha = 0.6, label = 't pdf')
```

output:

```
with df = 20 value of t_0 satisfying equation P(t < t_0) = 90 is 1.3253407069850462
with df = 15 value of t_0 satisfying equation P(t < t_0) = 95 is 1.7530503556925547
with df = 25 value of t_0 satisfying equation P(t < t_0) = 99 is 2.4851071754106413
with df = 30 value of t_0 satisfying equation P(|t| < t_0) = 90 is 1.6972608943617378
with df = 42 value of t_0 satisfying equation P(|t| < t_0) = 95 is 2.018081697095881
with df = 10 value of t_0 satisfying equation P(|t| < t_0) = 99 is 3.169272667175838
```

```
0.0 1.25 0.0 1.0
[<matplotlib.lines.Line2D at 0x7f5cacec1590>]
```



5. If F follows Snedecor's F distribution, then find

(i) $P(F < 1.5)$ with $df_1 = 5, df_2 = 14$ (ii) $P(F > 2.5)$ with $df_1 = 15, df_2 = 14$

(iii) $P(0.5 < F < 4.1)$ with $df_1 = 13, df_2 = 17$

Code:

```
from scipy.stats import f
P1 = f.cdf(1.5, 5, 14)
print("With df1 = 5, df2 = 14, P(F < 1.5) = " + str(P1))
P2 = f.sf(2.5, 15, 14)
print("With df1 = 15, df2 = 14, P(F > 2.5) = " + str(P2))
P3 = f.cdf(4.1, 13, 17) - f.cdf(0.5, 13, 17)
print("With df1 = 13, df2 = 17, P(0.5 < F < 4.1) = " + str(P3))
```

output:

1. With df1 = 5, df2 = 14, $P(F < 1.5) = 0.7480582329156877$
2. With df1 = 15, df2 = 14, $P(F > 2.5) = 0.047343688940604506$
3. With df1 = 13, df2 = 17, $P(0.5 < F < 4.1) = 0.8911203850553827$

6. If F follows Snedecor's F distribution, then find value of F_0 satisfying the given equation

(i) $P(F < F_0) = 0.90$ with $df_1 = 5, df_2 = 14$

(ii) $P(F < F_0) = 0.95$ with $df_1 = 15, df_2 = 13$

(iii) $P(F < F_0) = 0.99$ with $df_1 = 25, df_2 = 28$

Code:

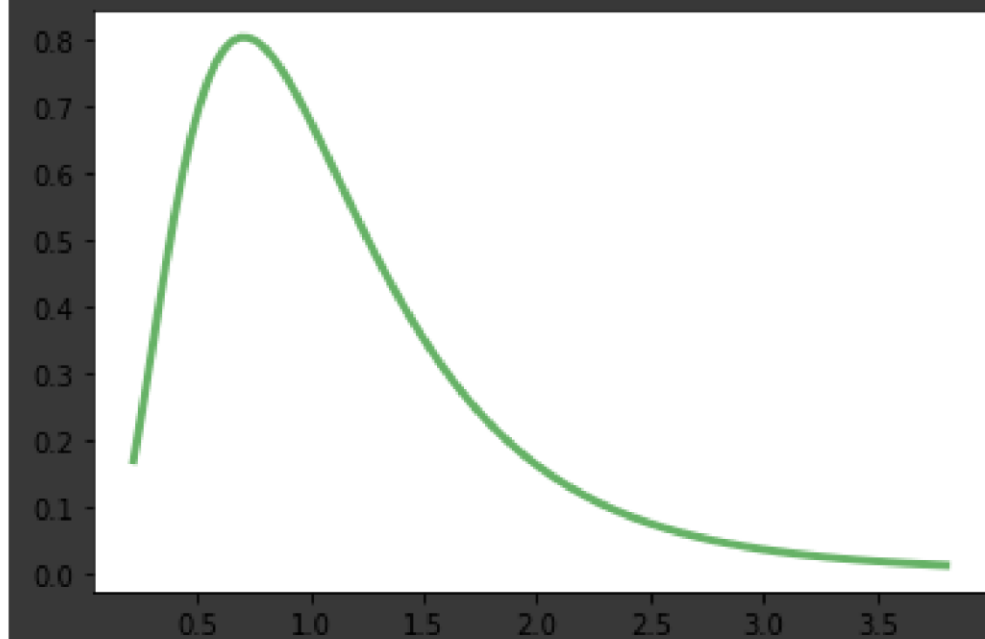

```
print("with df1 = 5, df2 = 14 value of f_0 satisfying equation P(F < F_0) = 0.90 is " + str(f.ppf(0.90, 5, 14)))
print("with df1 = 15, df2 = 13 value of f_0 satisfying equation P(F < F_0) = 0.95 is " + str(f.ppf(0.95, 15, 13)))
print("with df1 = 25, df2 = 28 value of f_0 satisfying equation P(F < F_0) = 0.99 is " + str(f.ppf(0.99, 25, 28)))
print("with df1 = 8, df2 = 29 value of F_0 satisfying equation P(F < F_0) = 0.99 is " + str(f.ppf(0.99, 8, 29)))

fig, ax = plt.subplots(1,1)
dfn, dfd = 10, 15
mean, var, skew, kurt = f.stats(dfn, dfd, moments='mvsk')
print(mean, var, skew, kurt)
x = np.linspace(f.ppf(0.01, dfn, dfd), f.ppf(0.99, dfn, dfd), 100)
ax.plot(x, f.pdf(x, dfn, dfd), 'g', lw=3, alpha = 0.6, label = 'F pdf')
```

output:

```
with df1 = 5, df2 = 14 value of f_0 satisfying equation P(F < F_0) = 0.90 is 2.3069430514007236
with df1 = 15, df2 = 13 value of f_0 satisfying equation P(F < F_0) = 0.95 is 2.533109983130745
with df1 = 25, df2 = 28 value of f_0 satisfying equation P(F < F_0) = 0.99 is 2.5060172667359417
with df1 = 8, df2 = 29 value of F_0 satisfying equation P(F < F_0) = 0.99 is 3.198218844688683
```

```
1.1538461538461537 0.556750941366326 2.268030280365984
11.634782608695652
[<matplotlib.lines.Line2D at 0x7f5caca2a1d0>]
```



7. If X follows χ^2 - distribution, then find

(i) $P(X < 1.5)$ with $df = 10$ (ii) $P(X > 2.5)$ with $df = 5$

(iii) $P(0.5 < X < 4.1)$ with $df = 2$

Code:

```
from scipy.stats import chi2
P1 = chi2.cdf(1.5, 10)
print("With df = 10, P(X < 1.5) = " + str(P1))
```

```
P2 = chi2.sf(2.5, 5)
print("With df = 5, P(X > 2.5) = " + str(P2)) P3 =
chi2.cdf(4.1, 2) - chi2.cdf(0.5, 2) print("With df
= 2, P(0.5 < X < 4.1) = " + str(P3))
```

Code:

output:

1. With df = 10, $P(X < 1.5) = 0.0010646777727857928$
2. With df = 5, $P(X > 2.5) = 0.7764950711233227$
3. With df = 2, $P(0.5 < X < 4.1) = 0.6500658794836006$

8. If X follows χ^2 - distribution, then find value of X_0 satisfying the given equation

- (i) $P(X < X_0) = 0.90$ with $df = 1$ (ii) $P(X < X_0) = 0.95$ with $df = 3$
- (iii) $P(X < X_0) = 0.99$ with $df = 2$ (iv) $P(X < X_0) = 0.05$ with $df = 1$
- (v) $P(X < X_0) = 0.025$ with $df = 3$ (vi) $P(X < X_0) = 0.005$ with $df = 2$

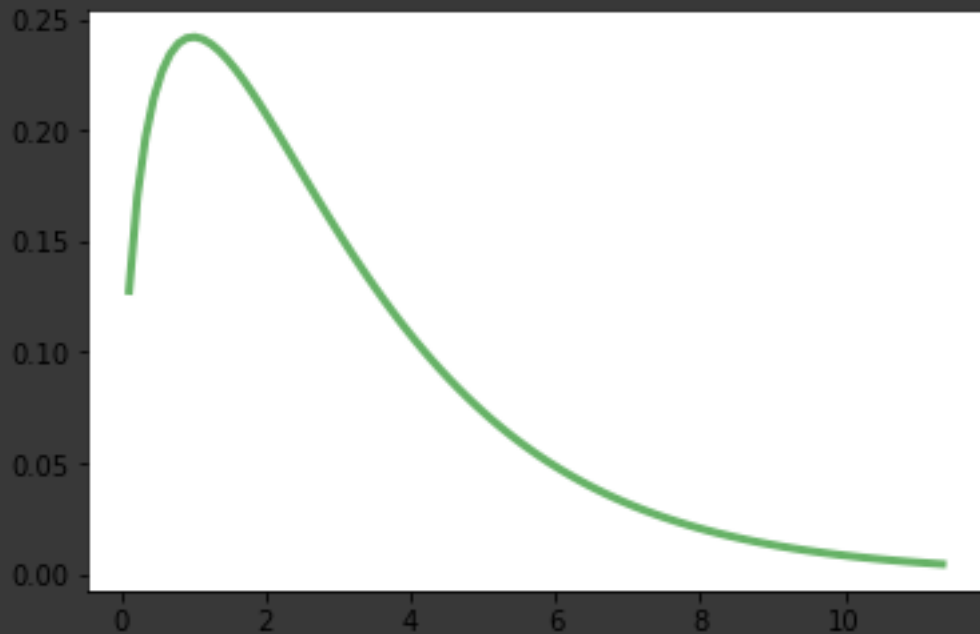
```
print("with df = 1 value of x_0 satisfying equation P(x < x_0) = 90 is " + str(chi2.ppf(0.90, 1))) print("with
df = 3 value of x_0 satisfying equation P(x < x_0) = 95 is " + str(chi2.ppf(0.95, 3))) print("with df = 2
value of x_0 satisfying equation P(x < x_0) = 99 is " + str(chi2.ppf(0.99, 2))) print("with df = 1 value of
x_0 satisfying equation P(x < x_0) = 0.05 is " + str(chi2.ppf(0.05, 1))) print("with df = 3 value of x_0
satisfying equation P(x < x_0) = 0.025 is " + str(chi2.ppf(0.025, 3))) print("with df = 2 value of x_0
satisfying equation P(x < x_0) = 0.005 is " + str(chi2.ppf(0.005, 2)))
```

```
fig,ax = plt.subplots(1,1) df = 3 mean, var, skew, kurt =
chi2.stats(df, moments ='mvsk') print(mean, var, skew,
kurt) x = np.linspace(chi2.ppf(0.01, df), chi2.ppf(0.99,
df),100)
ax.plot(x, chi2.pdf(x, df), 'g', lw=3, alpha = 0.6, label = 'chi2 pdf')
```

output:

```
with df = 1 value of x_0 satisfying equation P(x < x_0) = 90 is 2.705543454095404
with df = 3 value of x_0 satisfying equation P(x < x_0) = 95 is 7.814727903251179
with df = 2 value of x_0 satisfying equation P(x < x_0) = 99 is 9.210340371976184 with
df = 1 value of x_0 satisfying equation P(x < x_0) = 0.05 is
0.003932140000019522
with df = 3 value of x_0 satisfying equation P(x < x_0) = 0.025 is
0.21579528262389785
```

```
3.0 6.0 1.632993161855452 4.0
[<matplotlib.lines.Line2D at 0x7f5cac5b3490>]
```



9. Construct a 95 % confidence interval for population mean in an experiment that found the sample mean temperature for a certain city in August was 101.82, with a population standard deviation of 1.2. There were 6 samples in this experiment.

Code:

```
s_m = 101.82 #sample mean
pop_sd = 1.2 n = 6
#sample Size alpha =
0.05
z_0 = norm.ppf(1-alpha/2)
L = s_m - z_0*pop_sd/math.sqrt(n) U
= s_m + z_0*pop_sd/math.sqrt(n)
conf_interval = (L, U)
print("95% confidence interval for guiven population mean is "+ str(conf_interval))
```

output:

95%

10. Construct a 98% Confidence Interval for population mean based on the following data:

45,55,67,45,68,79,98,87,84,82.

```
confidence interval for given population mean is (100.85981766472894, 102.78018233527105)
```

Code:



```
from scipy.stats import t
data = [45, 55, 67, 45, 68, 79, 98, 87, 84, 82] s_m
= st.mean(data) #sample mean
s_sd = st.stdev(data) #sample standard deviation
n = len(data) #sampe size df = n - 1 alpha =
0.02
t_0 = t.ppf(1-alpha/2, df)
L = s_m - t_0*s_sd/math.sqrt(n)
U = s_m + t_0*s_sd/math.sqrt(n)
```




	<pre>conf_interval = (L, U) print("98% confidence interval for given population mean =" + str(conf_interval)) output: 98% confidence interval for given population mean =(54.78661680024991, 87.2133831997501)</pre> <p>11. 510 people applied to the Bachelor's in Elementary Education program at Florida State College. Of those applicants, 57 were men. Find the 90% CI of the true proportion of men who applied to the program.</p> <p>Code:</p> <pre>x = 57 n = 510 p_hat = x/n q_hat = 1 - p_hat alpha = 0.1 z_0 = norm.ppf(1-alpha/2) L = p_hat - z_0*math.sqrt(p_hat*q_hat/n) U = p_hat + z_0*math.sqrt(p_hat*q_hat/n) conf_interval = (L, U) print("98% confidence interval for guiven population proportion is "+ str(conf_interval)) output:</pre> <pre>98% confidence interval for guiven population proportion is (0.08881598300884237, 0.13471342875586353)</pre>
Conclusion	Different python and scipy libraries made it easy to calculate the probabilities of different distributions and vice versa.

Signature of Faculty