1. Aim:- Write a python program to find the best fit straight line and draw the scatter plot.

```
import numpy as np
import matplotlib.pyplot as plt
def mean(x):
    len=0
    sum=0
    for i in x:
        len+=1
        sum+=i
    return sum/len
x=np.array(list(map(float, input("Enter x values : ").split(",
"))))
y=np.array(list(map(float, input("Enter y values: ").split(",
"))))
x bar=mean(x)
y bar=mean(y)
num=0
den=0
length=len(x)
for i in range (length):
    num+=((x[i]-x bar)*(y[i]-y bar))
    den+=(x[i]-x bar)**2
slope=num/den
slope=slope.round(4)
print(f"slope={slope}")
y intercept=y bar-(slope*x bar)
y intercept=y intercept.round(4)
print(f"y intercept={y intercept}")
print(f"the required line is y={slope}x+{y intercept}")
y1=[]
for i in range(len(x)):
  y1.append(slope*x[i] + y intercept)
plt.scatter(y,x)
plt.scatter(y1,x)
plt.xlabel("X-axis")
plt.ylabel("y-axis")
plt.title("linear regression")
plt.plot(y,x)
plt.plot(y1,x,'r')
plt.show()
sse=0
```

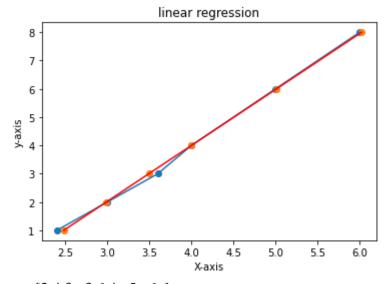
```
sst=0
ssr=0
for i in range(length):
    sse=sse+(y[i]-(slope*x[i]+y intercept))**2
    sst+=(y bar-y[i])**2
    ssr+=((slope*x[i]+y_intercept)-y_bar)**2
print("y = ",y)
for i in range(length):
    print(slope*x[i]+y intercept)
sse=sse.round(4)
ssr=ssr.round(4)
r square=1-(sse/sst)
r square=r square.round(4)
print(f"cost function of the line is {sse}")
print(f"The regression is {ssr}")
print(f"total error is {sst}")
print(f"The Goodness of fit model {r square}")
```

Output 1:-

Enter x values : 1,2,3,4,6,8 Enter y values : 2.4,3,3.6,4,5,6

slope=0.5059 y_intercept=1.9764

the required line is y=0.5059x+1.9764



y = [2.4 3. 3.6 4. 5. 6.] 2.4823 2.9882

3.4941

4.0

5.0118

6.0236

cost function of the line is 0.0188

The regression is 8.7018

total error is 8.72

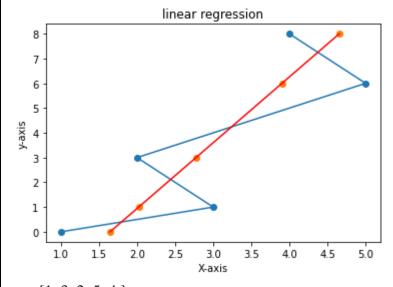
The Goodness of fit model 0.9978

Output 2:-

Enter x values : 0,1,3,6,8 Enter y values : 1,3,2,5,4

slope=0.3761 y_intercept=1.646

the required line is y=0.3761x+1.646



y = [1. 3. 2. 5. 4.]

1.646

2.0221

2.7742999999999998

3.9025999999999996

4.6548

cost function of the line is 3.6062

The regression is 6.3936

total error is 10.0

The Goodness of fit model 0.6394

2. Aim:- Write a python program to fit a second degree parabola of the form $y=a+bx+cx^2$ and draw the scatter plot.

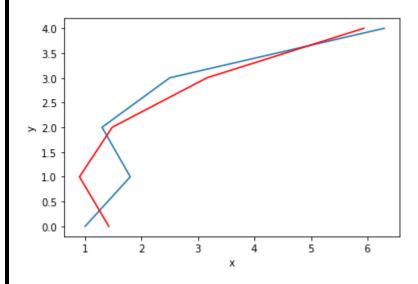
```
import numpy as np
import matplotlib.pyplot as plt
x=list(map(float, input().split(",")))
y=list(map(float, input().split(",")))
def mean(x):
    len=0
    sum=0
    for i in x:
        len+=1
        sum+=i
    return sum/len
xy=[]
x2y = []
x2 = []
x3 = []
\times 4 = []
for i in range(len(x)):
  xy.append(x[i]*y[i])
  x2y.append(x[i]*x[i]*y[i])
  x2.append(x[i]*x[i])
  x3.append(x2[i]*x[i])
  x4.append(x2[i]*x2[i])
\#print(x, y, xy, x2y, x2, x3, x4)
print("Equations to be solved are : ")
print("{} = {} a + {} b + {} c".format(sum(y), len(x), sum(x), sum(x))
print("{} = {} a + {} b + {} c".format(sum(xy), sum(x), sum(x2), sum
(x3))
print("{} = {} a + {} b + {} c".format(sum(x2y), sum(x2), sum(x3), s
um(x4))
def getMinor(m,i,j):
  return [p[:j]+p[j+1:] for p in m[:i]+m[i+1:]]
def getDeterminant(a):
  if len(a) == 1:
    return a[0][0]
  if len(a) == 2:
    return a[0][0]*a[1][1]-a[0][1]*a[1][0]
  det=0
  for c in range(len(a)):
```

```
det+=(-1)**c*a[0][c]*getDeterminant(getMinor(a,0,c))
  return det
d=[[len(x), sum(x), sum(x2)], [sum(x), sum(x2), sum(x3)], [sum(x2), sum(x3)]
um(x3), sum(x4)]
d1=[[sum(y), sum(xy), sum(x2y)], [sum(x), sum(x2), sum(x3)], [sum(x2)]
), sum(x3), sum(x4)]
d2=[[len(x), sum(x), sum(x2)], [sum(y), sum(xy), sum(x2y)], [sum(x2)]
, sum(x3), sum(x4)]]
d3=[[len(x), sum(x), sum(x2)], [sum(x), sum(x2), sum(x3)], [sum(y), sum(x3)]
um(xy), sum(x2y)]
a=getDeterminant(d1)/getDeterminant(d)
b=getDeterminant(d2)/getDeterminant(d)
c=getDeterminant(d3)/getDeterminant(d)
print()
print('a = {},b = {},c = {}'.format(a,b,c))
print('final equation is :')
print('y = {} + {}x + {}x^2'.format(a,b,c))
print()
y1=[]
for i in range(len(y)):
  y1.append(a + b*x[i] + c*x[i]*x[i])
plt.xlabel('x')
plt.ylabel('y')
plt.plot(y,x)
plt.plot(y1,x,'r')
plt.show()
Output 1:-
0,1,2,3,4
1,1.8,1.3,2.5,6.3
Equations to be solved are:
37.1 = 10.0a + 30.0b + 100.0c
130.3 = 30.0a + 100.0b + 354.0c
```

a = 1.42000000000000006, b = -1.070000000000116, c = 0.550000000000000005

 $y = 1.4200000000000006 + -1.07000000000116x + 0.5500000000000025x^2$

final equation is:



Output 2:-

2,4,6,8,10

3.07,12.85,31.47,57.38,91.29

Equations to be solved are:

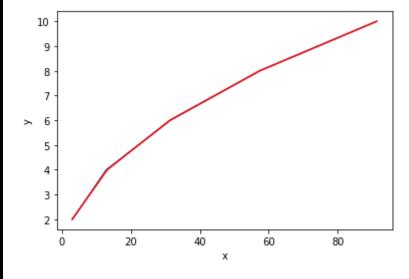
196.06 = 5a + 30.0b + 220.0c

1618.30000000000002 = 30.0a + 220.0b + 1800.0c

14152.12 = 220.0a + 1800.0b + 15664.0c

a=0.695999999996008, b=-0.8550714285710326, c=0.9919642857143065 final equation is :

 $y = 0.6959999999996008 + -0.8550714285710326x + 0.9919642857143065x^2$



3. Aim:- Write a python program to find Karl Pearson's correlation coefficient.

Source code:-

```
import numpy as np
import statistics as st
import math
x input=list(map(int,input("enter x values:").split()))
y input=list(map(int,input("enter y values:").split()))
x=np.array(x input)
y=np.array(y input)
print(x)
print(y)
xsum=sum(x)
ysum=sum(y)
xy=x*y
xysum=sum(xy)
xsquare=x**2
ysquare=y**2
xsquaresum=sum(xsquare)
ysquaresum=sum(ysquare)
xbar=st.mean(x input)
ybar=st.mean(y input)
cov xy=(xysum/np.size(x))-(xbar*ybar)
sigma x=math.sqrt((xsquaresum/np.size(x))-(xbar**2))
sigma y=math.sqrt((ysquaresum/np.size(y))-(ybar**2))
r=round(cov xy/(sigma x*sigma y),4)
r=round(cov xy/(sigma x*sigma y),4)
print('the correlation coefficient of given data is:',r)
r builtin=np.corrcoef(x, y)
print ('the correlation coefficient of given data by builtin f
unction is:',round(r builtin[1,0],4))
```

Output 1:-

enter x values: 3 7 4 2 0 4 1 2

```
enter y values:11 18 9 4 7 6 3 8
[3 7 4 2 0 4 1 2]
[11 18 9 4 7 6 3 8]
the correlation coefficient of given data is: 0.7867
the correlation coefficient of given data by builtin function is: 0.7867
```

Output 2:-

enter x values:35 39 45 53 63 75 89 105 117

enter y values:181 151 126 106 91 81 76 68 65

[35 39 45 53 63 75 89 105 117]

[181 151 126 106 91 81 76 68 65]

the correlation coefficient of given data is: -0.8834

the correlation coefficient of given data by builtin function is: -0.8834

4. Aim:- Write a python program to find the Spearman's correlation coefficient between x and y variables.

```
import numpy as np
cf=[]
def find rank(l):
    r={ }
    s=list(set(l))
    w=1.copy()
    s.sort(reverse=True)
    w.sort(reverse=True)
    a=1
    for i in s:
        m=l.count(i)
        if m==1:
            r[i]=a
            a+=1
        else:
            q=0
            for j in range(m):
                 q+=a
                 a+=1
            r[i]=q/m
            cf.append(m*(m**2-1)/12)
    return r
x=list(map(int, input().split(",")))
y=list(map(int, input().split(",")))
x rank=find rank(x)
y rank=find rank(y)
n=len(x)
d=[]
for i in range(n):
    d.append(x rank[x[i]]-y_rank[y[i]])
d = [i*i for i in d]
if len(cf)!=0:
    sum d 2=sum(d 2)+sum(cf)
else:
    sum d 2=sum(d 2)
print(d)
print(d 2)
print(x rank)
print(cf)
print(sum d 2)
num=6*sum d 2
den=n*(n**2 - 1)
print(num)
```

```
print(den)
rankcorelation=1 - num/den
print("Rank corelation is {:.5f}".format(rankcorelation))
```

Output 1:-

```
81,78,73,73,69,68,62,58
10,12,18,18,18,22,20,24
[-7, -5, -1.5, -1.5, 0.0, 4, 4, 7]
[49, 25, 2.25, 2.25, 0.0, 16, 16, 49]
{81: 1, 78: 2, 73: 3.5, 69: 5, 68: 6, 62: 7, 58: 8}
[0.5, 2.0]
162.0
972.0
504
Rank corelation is -0.92857
```

Output 2:-

```
68,64,75,50,64,80,75,40,55,64
62,58,68,45,81,60,68,48,50,70
[-1, -1.0, -1.0, -1, 5.0, -5, -1.0, 1, 0, 4.0]
[1, 1.0, 1.0, 1, 25.0, 25, 1.0, 1, 0, 16.0]
{80: 1, 75: 2.5, 68: 4, 64: 6.0, 55: 8, 50: 9, 40: 10}
[0.5, 2.0, 0.5]
75.0
450.0
990
Rank corelation is 0.54545
```

5. Aim:- Write a python program to classify the data based on one-way Anova.

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
%matplotlib inline
import scipy.stats as s
print("ANOVA ONE WAY CLASSIFICATION")
deg=float(input("Enter the degrees of freedom:"))
tn=int(input("Enter the no.of Treatments: "))
f list=[]
d=[]
for i in range(tn):
  l=[float(j) for j in input().split()]
  d.append(len(1))
  f list.append(l)
#print(f list)
ti=[]
for i in f list:
  ti.append(sum(i))
#print(ti)
ti sum=sum(ti)
#print(ti sum)
ti2 ni=[]
for i in range(len(ti)):
  ti2 ni.append((ti[i]**2)/d[i])
#print(ti2 ni)
#print(sum(ti2 ni))
rss=0.0
for i in f list:
  for j in range(len(i)):
    rss+=(i[j]**2)
cf = (ti sum**2) / sum(d)
sst=rss-cf
sstr=sum(ti2 ni)-cf
sse=sst-sstr
print("The row sum of squares: ",rss)
print("The correction factor:",cf)
print("sum of squares due to total:",sst)
print("sum of squares due to treatments:",sstr)
print("sum of squares due to error:",sse)
#print("sst:{},sstr:{},sse:{}".format(sst,sstr,sse));
#Constructing a dataframe
df=pd.DataFrame()
df["source of variation"]=["treatments", "errors", "total"]
df["sum of squares"]=[sstr,sse,sst]
```

```
x=(tn-1)
y=sum(d)-tn
z=sum(d)-1
df["Degrees of freedom"] = [x, y, z]
x1=sstr/x
y1=sse/y
df["Mean sum of Squares"] = [x1, y1, "*"]
f=x1/y1
df["Variance ratio"]=["*",f,"*"]
print(df)
if f<1:
  f=1/f
  tv=s.f.ppf(1-deg,y,x)
else:
  tv=s.f.ppf(1-deg,x,y)
print("f is :",f)
print("The table value is:",tv)
#Inference or decision
if f>=tv:
  print("Reject the Null Hypothesis")
else:
  print("Accept the NUll Hypothesis")
```

Output 1:-

```
ANOVA ONE WAY CLASSIFICATION
```

```
Enter the degrees of freedom:0.05 Enter the no.of Treatments: 3
```

13 10 8 11 8 13 11 14 14

4 1 3 4 2 4

The row sum of squares: 1262.0 The correction factor: 960.0 sum of squares due to total: 302.0

sum of squares due to treatments: 270.0

sum of squares due to error: 32.0

source of variation sum of squares $\,$ Degrees of freedom Mean sum of Squares $\,\setminus\,$

0	treatments	270.0	2	135.0
1	errors	32.0	12	2.666667
2	total	302.0	14	*

Variance ratio

0 *
1 50.625
2 *
f is: 50.625

The table value is: 3.8852938346523933

Reject the Null Hypothesis

Output 2:-

ANOVA ONE WAY CLASSIFICATION

Enter the degrees of freedom:0.05

Enter the no.of Treatments: 3

90 82 79 98 83 91

105 89 93 104 89 95 86

83 89 80 94

The row sum of squares: 138638.0 The correction factor: 137700.0 sum of squares due to total: 938.0

sum of squares due to treatments: 234.4523809523671 sum of squares due to error: 703.5476190476329

source of variation sum of squares Degrees of freedom Mean sum of Squares \

0 treatments 234.452381 2 117.22619 1 errors 703.547619 14 50.253401 2 total 938.000000 16 *

Variance ratio

0 *

1 2.332702

2 *

f is: 2.3327016142676427

The table value is: 3.738891832440735

Accept the NUll Hypothesis

6. Aim:- Write a python program to classify the data based on two-way Anova.

```
import scipy.stats as stats
print("Enter level of significance")
alpha=float(input())
print("Enter number of treatments")
t=int(input())
print("Enter number of blocks")
b=int(input())
1=[]
ti=[]
ti2ni=[]
for i in range(t):
    p=list(map(int,input().split()))
    ti.append(sum(p))
    ti2ni.append(sum(p)**2)
    1.append(p)
sti2=sum(ti2ni)
blocks=[]
squared blocks=[]
for i in range(b):
  for k in range(t):
    s+=l[k][i]
  blocks.append(s)
  squared blocks.append(s**2)
G=sum(blocks)
sbi2=sum(squared blocks)
m = []
for i in 1:
  m.extend(i)
rss=0
for i in m:
  rss+=i**2
cf=(G^**2) / len(m)
st2=rss-cf
str2=(sti2/b) - cf
sb2=(sbi2/t) - cf
se2=st2 - str2 - sb2
dft=t-1
dfb=b-1
dfe=dft*dfb
mst2=str2/dft
msb2=sb2/dfb
mse2=se2/dfe
ftr=mst2/mse2
```

```
fb=msb2/mse2
print("rss : {} ,cf : {} ,st2 : {} ,str2 : {} ,sb2 : {} ,se2 :
{} ,ftr : {} ,fb : {}".format(rss,cf,st2,str2,sb2,se2,ftr,fb)
if ftr<1.0:
  ftr=(1.0/ftr)
 t ftr=stats.f.ppf(1-alpha,dfe,dft)
 print("table value for treatments")
 print(t ftr)
  if ftr>t ftr:
    print("we reject treatments")
  else:
    print("we accept treatments")
else:
  t ftr=stats.f.ppf(1-alpha,dft,dfe)
 print("table value for treatments")
 print(t ftr)
  if ftr>t ftr:
    print("we reject treatments")
    print("we accept treatments")
if fb<1.0:
  fb = (1.0/fb)
  t fb=stats.f.ppf(1-alpha,dfe,dfb)
 print("Table value for blocks")
 print(t fb)
 if fb>t fb:
    print("we reject blocks")
  else:
    print("we accept blocks")
else:
 print("Table value for blocks")
 t fb=stats.f.ppf(1-alpha,dfb,dfe)
 print(t fb)
  if fb>t fb:
    print("we reject blocks")
  else:
    print("we accept blocks")
```

Output 1:-

```
Enter level of significance
0.05
Enter number of treatments
Enter number of blocks
13793
6631
11 5 15 5
rss: 786, cf: 588.0, st2: 198.0, str2: 56.0, sb2: 90.0, se2: 52.0, ftr: 3.230769230769231, fb
: 3.4615384615384617
table value for treatments
5.143252849784718
we accept treatments
Table value for blocks
4.757062663089414
we accept blocks
```

Output 2:-

```
Enter level of significance
0.01
Enter number of treatments
Enter number of blocks
45 43 51
47 46 52
48 50 55
42 37 49
rss: 26867,cf: 26602.0833333333333,st2: 264.916666666679,str2: 110.91666666666788
,sb2: 135.166666666666788,se2: 18.8333333333212,ftr: 11.77876106194779,fb:
21.530973451329015
table value for treatments
9.779538240923273
we reject treatments
Table value for blocks
10.92476650083833
we reject blocks
```

7. Aim:- Write a python program to fit a multiple regression model for any given data.

```
import scipy.stats as stats
def transportMatrix(a):
  res = [[a[j][i] for j in range(len(a))] for i in range(len(a
[0]))]
  return res
import numpy as np
no ind=int(input("Enter no of independent variables"))
indep=[]
for i in range (no ind):
  print(f"Enter the values for independent variable {i+1}")
  li=[float(x) for x in input().split()]
  indep.append(li)
print("Enter the values of output")
y=[float(x) for x in input().split()]
alpha=float(input("Enter level of significance"))
x trans=[]
x trans.append(np.ones(len(indep[0])))
for i in range(len(indep)):
  x trans.append(indep[i])
s1=np.dot(x trans, transportMatrix(x trans))
s2=np.linalq.inv(s1)
s3=np.dot(x trans,y)
s4=np.dot(s2,s3)
s4=s4.round(4)
y = [x.round(4) for x in s4]
y = qu = str(s4[0])
for i in range(1, no ind+1):
    y equ+='+'+str(s4[i])+'x'+str(i)
print()
print(f"The required multi linear regression is y={y equ}")
print()
y fit=[float(0) for i in range(len(indep[0]))]
for i in range(len(indep[0])):
    y fit[i]=s4[0]
    k=0
    for j in range(1, no ind+1):
        y fit[i]+=s4[j]*indep[k][i]
        k+=1
y fit=[x.round(4) for x in y fit]
print("Fitted values are")
print(y fit)
print()
sse=0
```

```
for i in range(len(y fit)):
    sse+=(y[i]-(y fit[i]))**2
sse=sse.round(4)
print(f"The sum of squares due to error is {sse}")
y mean=np.mean(y).round(4)
sst=0
for i in range(len(y fit)):
    sst+=(y[i]-y mean)**2
sst=sst.round(4)
print(f"The sum of squares due to total is {sst}")
ssr=sst-sse
print(f"The sum of squares due to total is {ssr}")
print()
r square=(ssr/sst).round(4)
print(f"The R^2 value is {r square}")
print()
print("By using R^2 test")
if(r square>=0.90):
    print("We accept the model")
else:
    print("We reject the model")
print()
m ssr=(ssr/no ind).round(4)
m sse=(sse/(len(y fit)-no ind-1))
f cal=(m ssr/m sse).round(4)
dfe=len(y fit)-no ind-1
print(f"The calucalted value of F is {f cal}")
f tab=stats.f.ppf(1-alpha,no ind,dfe).round(4)
print(f"The table value of F is {f tab}")
print()
print("By using ANOVA test")
if(f cal>f tab):
    print("We accept the model")
else:
    print("We reject the model")
print()
param value=[]
for i in range (no ind+1):
    temp=(s4[i]/((m sse*s2[i][i])**0.5)).round(4)
    param value.append(temp)
print("t values parameters are :")
print(param value)
t table=stats.t.ppf(1-alpha/2,len(y fit)-no ind-1)
print(f"The table value of t : {t table}")
i=0
for a in param value:
    if (abs(a))>t table:
        print("Beta"+str(i)+" is contributing to the model")
```

```
else:
     print("Beta"+str(i)+" is not contributing to the model
")
     i+=1
```

Output 1:-

Enter no of independent variables2
Enter the values for independent variable 1
9 8 7 14 12 10 7 4 6 5 7 6
Enter the values for independent variable 2
62 58 64 60 63 57 55 56 59 61 57 60
Enter the values of output
100 110 105 94 95 99 104 108 105 98 105 110
Enter level of significance.05

The required multi linear regression is y=133.4605+-1.2485x1+-0.351x2

Fitted values are

[100.462, 103.1145, 102.257, 94.9215, 96.3655, 100.9685, 105.416, 108.8105, 105.2605, 105.807, 104.714, 104.9095]

The sum of squares due to error is 151.4104 The sum of squares due to total is 330.25 The sum of squares due to total is 178.8396

The R^2 value is 0.5415

By using R^2 test We reject the model

The calucalted value of F is 5.3152 The table value of F is 4.2565

By using ANOVA test We accept the model

t values parameters are :

[5.0882, -2.8079, -0.7711] The table value of t : 2.2621571627409915

Beta0 is contributing to the model Beta1 is contributing to the model

Beta2 is not contributing to the model

Output 2:-

Enter no of independent variables2
Enter the values for independent variable 1
-5 -4 -1 2 2 3 3
Enter the values for independent variable 2
5 4 1 -3 -2 -2 -3
Enter the values of output
11 11 8 2 5 5 4
Enter level of significance0.05

The required multi linear regression is y=6.5714+1.0x1+2.0x2

Fitted values are [11.5714, 10.5714, 7.5714, 2.5714, 4.5714, 5.5714, 3.5714]

The sum of squares due to error is 1.7143 The sum of squares due to total is 73.7143 The sum of squares due to total is 72.0

The R^2 value is 0.9767

By using R^2 test We accept the model

The calucalted value of F is 83.9993 The table value of F is 6.9443

By using ANOVA test We accept the model

t values parameters are:
[26.5579, 2.1523, 4.3046]
The table value of t: 2.7764451051977987
Beta0 is contributing to the model
Beta1 is not contributing to the model
Beta2 is contributing to the model

8. Aim:- Write a python program to fit a multivariate regression model for any given data.

```
import numpy as np
q=int(input('Enter no. of dependent variables:'))
yt=[]
for i in range(q):
  print('Enter y'+str(i+1),'data')
  yt.append([float(j) for j in input().split()])
n=len(yt[0])
p=int(input('Enter no. of independent variables:'))
xt=[]
xt.append([1]*n)
for i in range(p):
  print('Enter x'+str(i+1),'data')
  xt.append([float(j) for j in input().split()])
p=p+1
g=float(input('Enter level of significance: '))
def transpose(m):
    return [[m[j][i] for j in range(len(m))] for i in range(le
n (m[0]))]
def getMinor(m,i,j):
  return [p[:j]+p[j+1:] for p in m[:i]+m[i+1:]]
def getDeterminant(a):
  if len(a) == 1:
    return a[0][0]
  if len(a) == 2:
    return a[0][0]*a[1][1]-a[0][1]*a[1][0]
  det=0
  for c in range(len(a)):
    det+=(-1)**c*a[0][c]*getDeterminant(getMinor(a,0,c))
  return det
def coFactor(m):
  return [[(-
1) **(i+j) *getDeterminant(getMinor(m,i,j)) for j in range(len(m)
)] for i in range(len(m))]
def Inverse(m):
  det=getDeterminant(m)
  if det==0: print('det=0,Inverse not posssible');exit()
```

```
adj=transpose(coFactor(m))
  return [[adj[i][j]/det for j in range(len(adj[0]))] for i in
 range(len(adj))]
def Multiply(a,b):
    r1, c1, r2, c2 = len(a), len(a[0]), len(b), len(b[0])
    if c1!=r2: print('Multiplication not possible');exit()
    mm = []
    for i in range (0, r1):
        C = []
        for j in range (0,c2):
            p=0
            for k in range (0, c1):
                p=p+(a[i][k]*b[k][j])
            c.append(p)
        mm.append(c)
    return mm
x=transpose(xt)
y=transpose(yt)
xtxi=Inverse(Multiply(xt,x))
xty=Multiply(xt,y)
b=Multiply(xtxi,xty)
yt=np.array(yt)
xt=np.array(xt)
for i in range(q):
 print('\nFOR THE EQUATION y'+str(i+1),':')
  ye='y'+str(i+1)+' = '
  ye=ye+str('%.3f'%b[0][i])+' + '
  y1=np.array(b[0][i]*xt[0])
  f=0
  for j in range (1,p):
      ye=ye+str('(%.3f)'%b[j][i])+'x'+str(j);y1=y1+(b[j][i]*xt
[j])
      if j!=p-1:ye=ye+' + '
  print('Model fit equation is ',ye)
  sse=np.sum((yt[i]-y1)**2)
  sst=np.sum((yt[i]-np.mean(yt[i]))**2)
  ssr=sst-sse
  print('\nUsing Coefficient of Determination Method: ')
  r2=ssr/sst
  print('R^2 = %.5f'%(r2))
  if(r2>0.90):print('Model fits good (R^2>0.90)')
  else:print('Model does not fit good (R^2<=0.90)')
  print('\nUSing ANOVA Test:')
```

```
mssr=ssr/(p-1)
msse=sse/(n-p)
f=mssr/msse
import scipy.stats as s
if f>=1: ft=s.f.ppf(1-g,p-1,n-p)
else: f=1/f; ft=s.f.ppf(1-q,n-p,p-1)
print('Calculated value of f is %.5f'%(f))
print('Table value of f is %.5f'%(ft))
if f>ft: print('Model fits good')
else: print('Model does not fit good')
print('\nTest of Individual Parameters:')
cjj=[xtxi[j][j] for j in range(len(xtxi))]
tt=s.t.ppf(1-g/2,n-p)
print('Table value of t is %.5f'%(tt))
for j in range(p):
    e=np.sqrt(msse*cjj[j])
    t=b[j][i]/e
    q='b['+str(j)+']'
    print('t(cal) for',q,'is %.5f'%(t))
    if abs(t)>tt: print(q,'is contributing to the model')
    else: print(q,'is not contributing to the model')
```

Output 1:-

```
Enter no. of dependent variables:2
Enter v1 data
10 12 11 9 9 10 11 12 11 10 11 12
Enter y2 data
100 110 105 94 95 99 104 108 105 98 103 110
Enter no. of independent variables:3
Enter x1 data
987141210746576
Enter x2 data
62 58 64 60 63 57 55 56 59 61 57 60
Enter x3 data
1 1.3 1.2 0.8 0.8 0.9 1 1.2 1.1 1.0 1.2 1.2
Enter level of significance: 0.05
FOR THE EQUATION y1:
Model fit equation is y1 = 10.897 + (-0.045)x1 + (-0.088)x2 + (5.035)x3
Using Coefficient of Determination Method:
R^2 = 0.92380
Model fits good (R^2>0.90)
USing ANOVA Test:
Calculated value of f is 32.32738
Table value of f is 4.06618
```

Model fits good

Test of Individual Parameters:
Table value of t is 2.30600
t(cal) for b[0] is 4.23735
b[0] is contributing to the model
t(cal) for b[1] is -0.82835
b[1] is not contributing to the model
t(cal) for b[2] is -2.27524
b[2] is not contributing to the model
t(cal) for b[3] is 5.46181
b[3] is contributing to the model

FOR THE EQUATION y2:

Model fit equation is y2 = 91.097 + (-0.064)x1 + (-0.294)x2 + (27.835)x3

Using Coefficient of Determination Method: $R^2 = 0.86551$

Model does not fit good (R^2<=0.90)

USing ANOVA Test:

Calculated value of f is 17.16130 Table value of f is 4.06618 Model fits good

Test of Individual Parameters:
Table value of t is 2.30600
t(cal) for b[0] is 5.26478
b[0] is contributing to the model
t(cal) for b[1] is -0.17534
b[1] is not contributing to the model
t(cal) for b[2] is -1.13500
b[2] is not contributing to the model
t(cal) for b[3] is 4.48725
b[3] is contributing to the model

9. Aim:- Write a python program to classify the treatments based on MANOVA Test.

```
import numpy as np
import scipy.stats as s
k = int(input("Enter no.of treatments : "))
p = int(input("Enter no.of observations in each subgroup : "))
T = []
n = []
alp = float(input('Enter level of significance : '))
for i in range(k):
    n.append(int(input("Enter no.of sub groups in treatment :"
)))
    t = []
    for j in range(n[i]):
        a = list(map(int, input("Enter rowwise: ").split()))
        a = np.matrix(a).reshape(p, 1)
        t.append(a)
    T.append(t)
print()
for i in range(k):
    print()
    print("Treatment ",i+1)
    print(T[i])
print()
yi_ = []
y = 0
for i in range(k):
    yi .append(sum(T[i])/n[i])
    y += sum(T[i])
y_{\underline{}} = y_{\underline{}}/sum(n)
print("yi_ = ",yi_)
print()
print("y__ = ",y__)
print()
# For y1
ssey1 = 0
ssty1 = 0
for i in range(k):
    for j in range(n[i]):
        ssey1 = ssey1 + (float(T[i][j][0]) -
float(yi [i][0]))**2
        ssty1 = ssty1 + (float(T[i][j][0]) - float(y [0])) **2
```

```
ssry1 = ssty1-ssey1
print("ssey1 = ",ssey1)
print("ssty1 = ",ssty1)
print("ssrt1 = ",ssry1)
print()
# For y2
ssey2 = 0
ssty2 = 0
for i in range(k):
    for j in range(n[i]):
        ssey2 = ssey2 + (float(T[i][j][1]) -
float(yi [i][1]))**2
        ssty2 = ssty2 + (float(T[i][j][1]) - float(y [1])) **2
ssry2 = ssty2-ssey2
print("ssey2 = ",ssey2)
print("ssty2 = ",ssty2)
print("ssry2 = ",ssry2)
print()
# For y1 and y2
ssey12 = 0
ssty12 = 0
for i in range(k):
    for j in range(n[i]):
        ssey12 = ssey12 + float((T[i][j][0]*T[i][j][1]) -
 (yi_[i][0]*yi [i][1]))
        ssty12 = ssty12 + float((T[i][j][0]*T[i][j][1]) -
 (y [0]*y [1]))
ssry12 = ssty12-ssey12
print("ssey12 = ",ssey12)
print("ssty12 = ",ssty12)
print("ssry12 = ", ssry12)
print()
B = np.matrix([[ssry1, ssry12], [ssry12, ssry2]])
W = np.matrix([[ssey1, ssey12], [ssey12, ssey2]])
T = np.matrix([[ssty1, ssty12], [ssty12, ssty2]])
print("B = ",B)
print()
print("W = ",W)
print()
print("T = ",T)
print()
def determinent(matrix):
```

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```
no rows=len(matrix)
    for row in matrix:
        if len(row)!=no rows:
            print("not a square matrix")
            return None
    if len(matrix) == 2:
        simple det=(matrix[0][0]*matrix[1][1])-
(matrix[1][0]*matrix[0][1])
        return simple det
    else:
        #cofactor expression
        answer=0
        no columns=no rows
        for j in range(no columns):
            cofactor=((-
1) ** (0+j) * matrix[0][j]) * (determinent(smallermatrix(matri
x, 0, j)))
            answer+=cofactor
        return answer
det w = determinent(np.array(W))
det t = determinent(np.array(T))
LAMBDA = float(det w/det t)
print("LAMBDA*=",LAMBDA)
F = ((sum(n)-k-1)/(k-1))*((1-(LAMBDA)**.5)/(LAMBDA**.5))
alp = .05
cou = 0
if F<1:
    cou=1
    F=1/F
print("F cal=",F)
if cou==1:
    Ft=s.f.ppf(1-alp,p*(sum(n)-k-1),p*(k-1))
else:
    Ft=s.f.ppf(1-alp,p*(k-1),p*(sum(n)-k-1))
print("F t.v=",Ft)
if F>Ft:
   print("We reject Ho")
    print("There is no homoginity among the table")
    print("We reject the model")
else:
    print("We accept Ho")
    print("There is homoginity among the table")
    print("We accept the model")
```

Output 1:-

```
Enter no. of treatments: 3
Enter no. of observations in each subgroup: 2
Enter level of significance: 0.05
Enter no. of sub groups in treatment :4
Enter rowwise: 23
Enter rowwise: 3 4
Enter rowwise: 5 4
Enter rowwise: 25
Enter no. of sub groups in treatment :3
Enter rowwise: 48
Enter rowwise: 5 6
Enter rowwise: 67
Enter no. of sub groups in treatment :5
Enter rowwise: 7 6
Enter rowwise: 87
Enter rowwise: 108
Enter rowwise: 9 5
Enter rowwise: 7 6
Treatment 1
[matrix([[2],
     [3]]), matrix([[3],
     [4]]), matrix([[5],
     [4]]), matrix([[2],
     [5]])]
Treatment 2
[matrix([[4],
     [8]]), matrix([[5],
     [6]]), matrix([[6],
     [7]])]
Treatment 3
[matrix([[7],
     [6]]), matrix([[8],
     [7]]), matrix([[10],
     [ 8]]), matrix([[9],
     [5]]), matrix([[7],
     [6]])]
yi_ = [matrix([[3.],
     [4.]]), matrix([[5.],
     [7.]]), matrix([[8.2],
     [6.4]])]
y_{\underline{}} = [[5.66666667]]
[5.75]
         11
```

```
ssty1 = 76.6666666666667
ssrt1 = 61.86666666666674
ssey2 = 9.2
ssty2 = 28.25
ssry2 = 19.05
ssey12 = 1.60000000000000156
ssty12 = 25.99999999999943
ssry12 = 24.39999999999928
B = [[61.8666666724.4]]
[24.4
         19.05
                - 11
W = [[14.8 \ 1.6]]
[ 1.6 9.2]]
T = [76.6666666726.
                         ]
[26.
         28.25
               ]]
LAMBDA*= 0.08967446023045059
F cal= 9.357513005519227
F t.v= 3.0069172799243438
We reject Ho
There is no homoginity among the table
We reject the model
Output 2:-
Enter no. of treatments: 3
Enter no. of observations in each subgroup: 2
Enter level of significance: 0.05
Enter no. of sub groups in treatment :3
Enter rowwise: 93
Enter rowwise: 62
Enter rowwise: 97
Enter no. of sub groups in treatment :2
Enter rowwise: 04
Enter rowwise: 20
Enter no. of sub groups in treatment :3
Enter rowwise: 3 8
Enter rowwise: 19
Enter rowwise: 27
Treatment 1
[matrix([[9],
    [3]]), matrix([[6],
    [2]]), matrix([[9],
```

```
[7]])]
Treatment 2
[matrix([[0],
     [4]]), matrix([[2],
    [([[0]
Treatment 3
[matrix([[3],
     [8]]), matrix([[1],
    [9]]), matrix([[2],
    [7]])]
yi_= [matrix([[8.],
    [4.]]), matrix([[1.],
    [2.]]), matrix([[2.],
    [8.]])]
y_{\underline{}} = [[4.]]
[5.]]
ssey1 = 10.0
ssty1 = 88.0
ssrt1 = 78.0
ssey2 = 24.0
ssty2 = 72.0
ssry2 = 48.0
ssey12 = 1.0
ssty12 = -11.0
ssry12 = -12.0
B = [[78. -12.]
[-12. 48.]]
W = [[10. 1.]]
[ 1. 24.]]
T = [[88. -11.]]
[-11. 72.]]
LAMBDA*= 0.03845534995977474
F cal= 8.198859563778374
F t.v= 3.837853354555897
We reject Ho
There is no homoginity among the table
We reject the model
```

10. Aim:- Write a python program to classify the given observations using Linear Discriminant Analysis.

```
import numpy as np
import math
#11=[2.95, 2.53, 3.57, 3.16, 2.58, 2.16, 3.27]
#12=[6.63,7.79,5.65,5.47,4.46,6.22,3.52]
k = [[2.81, 5.46]]
11=[1300,1260,1220,1180,1060,1140,1100,1020,980,940]
12=[2.7,3.7,2.9,2.5,3.9,2.1,3.5,3.3,2.3,3.1]
X = []
for i in range(len(l1)):
  x.append([11[i], 12[i]])
print("x=",x)
x1 = []
x2 = []
#y=[1,1,1,1,0,0,0]
y=[1,1,1,1,1,0,0,0,0,0]
for i in range(len(y)):
  if(y[i] == y[0]):
    x1.append([11[i], 12[i]])
  else:
    x2.append([11[i], 12[i]])
print("x1=",x1)
print("x2=",x2)
mu = []
sum1=0
sum2=0
for i in range(len(l1)):
  sum1=sum1+l1[i]
  sum2=sum2+12[i]
mu.append([(sum1/len(l1)), (sum2/len(l2))])
print("Mean of x=", mu)
mu1=[]
sum3=0
for i in range (len(x1)):
  sum3=sum3+x1[i][0]
  sum4=sum4+x1[i][1]
mu1.append([(sum3/len(x1)),(sum4/len(x1))])
print("Mean of x1=", mu1)
mu2 = []
sum5=0
sum6=0
for i in range (len(x2)):
  sum5=sum5+x2[i][0]
  sum6=sum6+x2[i][1]
```

```
mu2.append([(sum5/len(x2)),(sum6/len(x2))])
print("Mean of x2=", mu2)
def transpose(x):
    xt = [[0 \text{ for i in range(len}(x))] \text{ for j in range(len}(x[0]))]
    for i in range(len(x)):
        for j in range(len(x[0])):
              xt[j][i]=x[i][j]
    return xt
def mul(x,y):
    res = [[0 \text{ for i in range}(len(y[0]))] \text{ for j in range}(len(x))]
))]
    for i in range (len(x)):
        for j in range(len(y[0])):
             for k in range(len(y)):
                 res[i][j] += x[i][k] * y[k][j]
    return res
a=[]
for i in range(len(x)):
  a.append([x[i][0]-mu[0][0],x[i][1]-mu[0][1]])
print("x-mu=",a)
for i in range(len(a)):
  c=mul(transpose(a),a)
for i in range(len(c)):
  for j in range(len(a[0])):
    c[i][j]=c[i][j]/len(a)
print("c=",c)
cin=np.linalg.inv(c)
print("c inverse=",cin)
f1 = 0
f2 = 0
f1=mul(mul(mul(mul,cin),transpose(k))[0][0]-
((mul(mul(mu1,cin),transpose(mu1))[0][0])*0.5)+math.log(len(x1
)/len(x))
print("F1=", f1)
f2=mul(mul(mu2,cin),transpose(k))[0][0]-
((mul(mul(mu2,cin),transpose(mu2))[0][0])*0.5)+math.log(len(x2
)/len(x))
print("F2=",f2)
if(f1>f2):
  print('The new observation is kept into first category i.e,p
assed category')
else:
  print('The new observation is kept into second category')
```

Output 1:-

```
x = [[2.95, 6.63], [2.53, 7.79], [3.57, 5.65], [3.16, 5.47], [2.58, 4.46], [2.16, 6.22], [3.27, 1.27]
3.52]]
x1 = [[2.95, 6.63], [2.53, 7.79], [3.57, 5.65], [3.16, 5.47]]
x2=[[2.58, 4.46], [2.16, 6.22], [3.27, 3.52]]
Mean of x = [[2.888571428571429, 5.677142857142857]]
Mean of x1 = [[3.0525, 6.385]]
Mean of x2 = [[2.67, 4.7333333333333333]]
x-mu= [[0.06142857142857139, 0.9528571428571428], [-0.358571428571429,
2.112857142857143], [0.681428571428571, -0.02714285714285669],
[0.27142857142857135, -0.2071428571428573], [-0.3085714285714287, -0.2071428571428573], [-0.3085714285714287, -0.2071428573], [-0.3085714285714287, -0.2071428573], [-0.3085714285714287, -0.2071428573], [-0.3085714285714287, -0.2071428573], [-0.3085714285714287, -0.2071428573], [-0.3085714285714287, -0.2071428573], [-0.3085714285714287, -0.2071428573], [-0.3085714285714287, -0.2071428573], [-0.3085714285714287, -0.2071428573], [-0.3085714285714287, -0.207142857], [-0.3085714285714287, -0.207142857], [-0.30857142857], [-0.30857142857], [-0.30857142857], [-0.30857142857], [-0.30857142857], [-0.30857142857], [-0.30857142857], [-0.30857142857], [-0.30857142857], [-0.30857142857], [-0.30857142857], [-0.30857142857], [-0.30857142857], [-0.30857142857], [-0.30857142857], [-0.30857142857], [-0.30857142857], [-0.30857142857], [-0.30857142857], [-0.30857142857], [-0.30857142857], [-0.30857142857], [-0.30857142857], [-0.30857142857], [-0.30857142857], [-0.30857142857], [-0.30857142857], [-0.30857142857], [-0.30857142857], [-0.30857142857], [-0.30857142857], [-0.30857142857], [-0.30857142857], [-0.30857142857], [-0.30857142857], [-0.30857142857], [-0.30857142857], [-0.30857], [-0.30857], [-0.30857], [-0.30857], [-0.30857], [-0.30857], [-0.30857], [-0.30857], [-0.30857], [-0.30857], [-0.30857], [-0.30857], [-0.30857], [-0.30857], [-0.30857], [-0.30857], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], [-0.3085], 
1.217142857142857], [-0.7285714285714286, 0.5428571428571427],
[0.38142857142857123, -2.157142857142857]]
c = [[0.2059836734693877, -0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.23093265306122449], [-0.2309326530612449], [-0.2309326530612449], [-0.2309326530612449], [-0.2309326530612449], [-0.2309326530612449], [-0.2309326530612449], [-0.2309326530612449], [-0.2309326530612449], [-0.2309326530612449], [-0.23093667], [-0.23093667], [-0.23093667], [-0.2309367], [-0.2309367], [-0.2309367], [-0.2309367], [-0.2309367], [-0.2309367], [-0.2309367], [-0.2309367], [-0.2309367], [-0.2309367], [-0.2309367], [-0.2309367], [-0.2309367], [-0.2309367], [-0.2309367], [-0.2309367], [-0.2309367], [-0.23097], [-0.23097], [-0.23097], [-0.23097], [-0.23097], [-0.23097], [-0.23097], [-0.23097], [-0.2309], [-0.2309], [-0.2309], [-0.2309], [-0.2309], [-0.2309], [-0.2309], [-0.2309], [
1.6921632653061225]]
c inverse= [[5.73171449 0.78221769]
 [0.78221769 0.69771022]]
F1= 43.82818099216788
F2= 43.86302017821443
The new observation is kept into second category
Output 2:-
x = [[1300, 2.7], [1260, 3.7], [1220, 2.9], [1180, 2.5], [1060, 3.9], [1140, 2.1], [1100, 3.5],
[1020, 3.3], [980, 2.3], [940, 3.1]]
x1 = [[1300, 2.7], [1260, 3.7], [1220, 2.9], [1180, 2.5], [1060, 3.9]]
x2 = [[1140, 2.1], [1100, 3.5], [1020, 3.3], [980, 2.3], [940, 3.1]]
Mean of x = [[1120.0, 3.000000000000000000]]
Mean of x1 = [[1204.0, 3.14]]
Mean of x2 = [[1036.0, 2.86]]
x-mu= [[180.0, -0.30000000000000027], [140.0, 0.699999999999997], [100.0, -
140.0, -0.70000000000000006], [-180.0, 0.09999999999999964]]
c = [[13200.0, -2.7999999999991], [-2.7999999999999, 0.329999999999999]]
c inverse= [[7.58941713e-05 6.43950545e-04]
 [6.43950545e-04 3.03576685e+00]]
F1= -16.559987701607927
F2= -4.471300403992386
The new observation is kept into second category
```

11. Aim:- Write a python program to find Principle components for the given variables.

```
import numpy as np
import pandas as pd
import scipy.stats as st
import copy
xn=int(input('Enter number of components='))
X = []
n=int(input("Enter no of observations="))
for i in range(xn):
    l=list(map(float,input().split()))
    x.append(1)
x copy=np.transpose(x)
print("x=",x_copy)
def mu cal(x):
  n=len(x)
  l=len(x[0])
  mu = []
  for i in range(1):
    s=0
    for j in range(n):
      s+=x[j][i]
    s=s/n
    mu.append(s)
  return mu
mu x = (mu cal(x copy))
print("Mean of x=", mu x)
xsubm=np.subtract(x copy,mu x)
print("\nx-mu=",xsubm)
#calculating pooled variance c inverse
pooled matrix=(np.dot(np.transpose(xsubm),xsubm))/n
print("pooled matrix=",pooled matrix)
# pooled matrix inv=np.linalg.inv(pooled matrix)
# print("pooled matrix inverse is:",pooled matrix inv)
eig value,eig vector=np.linalg.eig(pooled matrix)
eig value=np.sort(eig value)[::-1]
print("Eigen values=",eig value)
df=pd.DataFrame()
df['pricipal components']=[i+1 for i in range(len(eig value))]
df['variance explained']=[i for i in eig value]
sum=0
```

```
tot sum=np.sum(eig value)
1=[]
for i in eig value:
  sum+=i
  1.append((sum/tot sum) *100)
df['cumulative tot var']=1
print(df)
thres value=float(input("Enter The Threshold Value For Princip
al Component Analysis="))
1 = 0
#del df['cumulative tot var']
for i in df['cumulative tot var']:
  if i>=thres value:
    break
  else:
    1+=1
df=df.loc[0:1]
print(df)
eig values=list(df['variance explained'])
print ("Eigen values taken for to retain the principal componen
ts[based on the thershold is fixed]=",eig values)
co related coeff=[]
for m in eig values:
  l=copy.copy(pooled matrix)
  for i in range(len(l)):
    for j in range(len(l[0])):
      if i==j:
        l[i][j]-=m
  co related coeff.append(1)
co related coeff
def normalization(x):
  if len(x) == 2:
    total=(x[0][0])**2+(x[0][1])**2
    total=total**0.5
   # print(total)
   # 1=[]
    l=[-x[0][1]/total, x[0][0]/total]
   # l=list(-x[0][1]/total,x[0][0]/total)
    return 1
  else:
    1=[]
    copy1=list(copy.copy(x[0]))
    copy2=list(copy.copy(x[1]))
    copy1.append(x[0][0])
```

```
copy1.append(x[0][1])
    copy2.append(x[1][0])
    copy2.append(x[1][1])
    for i in range (1, len(copy1)-1):
      1.append(copy1[i]*copy2[i+1]-copy1[i+1]*copy2[i])
    sum=0
    for i in range(len(l)):
      sum+=1[i]**2
    sum = (sum * * 0.5) * * -1
    l=[i*sum for i in 1]
    return l
z=[normalization(i) for i in co related coeff]
print("The eigen vectors of z1 and z2 are:",z)
print("The Principal component equation are:")
for i in range(len(z)):
  print("\nZ", i+1, "=", end='')
  for j in range(len(z[0])):
    if j==len(z[0])-1:
      print(z[i][j],'X',j+1)
    else:
      print(z[i][j],"X",j+1,"+",end='')
corelated data=[]
for i in z:
  1=[]
  for j in range(len(x copy)):
    sum=0
    for k in range(len(x copy[j])):
      sum+=x copy[j][k]*i[k]
    1.append(sum)
  corelated data.append(1)
corelated data
corelated data df=pd.DataFrame()
for i in range(len(corelated data)):
  corelated data df[f'Z{i+1}']=corelated data[i]
corelated data df
```

Output 1:-

```
Enter number of components=2
Enter no of observations=4
2 1 0 -1
4 3 1 0.5
x= [[ 2. 4. ]
```

```
[1. 3.]
[0. 1.]
[-1. 0.5]]
Mean of x = [0.5, 2.125]
x-mu = [[1.5 \ 1.875]]
[0.5 \quad 0.875]
[-0.5 -1.125]
[-1.5 -1.625]]
pooled matrix= [[1.25]
                      1.5625
[1.5625 2.046875]]
Eigen values= [3.26093826 0.03593674]
 pricipal components variance explained cumulative tot var
0
             1
                     3.260938
                                    98.909976
1
             2
                     0.035937
                                    100.000000
```

Enter The Threshold Value For Principal Component Analysis=99 pricipal components variance explained cumulative tot var

0 1 3.260938 98.909976 1 2 0.035937 100.000000

Eigen values taken for to retain the principal components[based on the thershold is fixed]= [3.2609382570250163, 0.03593674297498417]

The eigen vectors of z1 and z2 are: [[-0.6135581035026774, -0.789649576474399], [-0.7896495764743992, 0.6135581035026774]]

The Principal component equation are:

Z 1 =-0.6135581035026774 X 1 +-0.789649576474399 X 2

Z 2 =-0.7896495764743992 X 1 +0.6135581035026774 X 2

Z1 Z2

0 -4.385715 0.874933

1 -2.982507 1.051025

2 -0.789650 0.613558

3 0.218733 1.096429

Output 2:-

```
Enter number of components=3
Enter no of observations=5
90 90 60 60 30
60 90 60 60 30
90 30 60 90 30
x = [[90. 60. 90.]]
[90. 90. 30.]
[60. 60. 60.]
[60. 60. 90.]
[30. 30. 30.]]
Mean of x = [66.0, 60.0, 60.0]
x-mu = [[24. 0.30.]]
[ 24. 30. -30.]
[-6. 0. 0.]
[-6. 0. 30.]
[-36. -30. -30.]]
pooled matrix= [[504. 360. 180.]
[360. 360. 0.]
[180. 0.720.]]
Eigen values= [910.06995304 629.11038668 44.81966028]
 pricipal components variance explained cumulative tot var
             1
                    910.069953
0
                                     57.453911
            2
                    629.110387
                                     97.170476
1
            3
                    44.819660
                                    100.000000
Enter The Threshold Value For Principal Component Analysis=95
 pricipal components variance explained cumulative tot var
             1
                    910.069953
                                     57.453911
0
                    629.110387
                                     97.170476
Eigen values taken for to retain the principal components[based on the thershold is fixed]=
[910.0699530410365, 629.1103866763252]
The eigen vectors of z1 and z2 are: [[0.6558022549801461, 0.42919779654868706,
0.6210576895914801], [0.38599879538810006, 0.5163664177215531, -
```

The Principal component equation are:

0.7644413990675454]]

Z 1 =0.6558022549801461 X 1 +0.42919779654868706 X 2 +0.6210576895914801 X 3

Z 2 =0.38599879538810006 X 1 +0.5163664177215531 X 2 +-0.7644413990675454 X 3

Z1 Z2

Z1 Z2

140.669263 -3.077849

116.281735 58.279627

102.363464 8.275429

120.995195 -14.657813

51.181732 4.137714

12. Aim:- Write a python program to group the given variables using Factor Analysis.

```
import numpy as np
n=int(input("enter no of variables"))
X = []
for i in range(n):
  p=[int(x) for x in input().split(" ")]
  x.append(p)
print(x)
mu = []
for i in range(n):
  mu.append(np.mean(x[i],axis=0))
print(mu)
from math import sqrt
from statistics import mean
n1=len(x[0])
si=[]
x=np.array(x)
for i in range(n):
  k = sum((x[i] - mu[i]) **2)
  si.append(round(sqrt(k/(n1-1)),4))
print(si)
a=[]
for i in range(n):
  a.append((x[i]-mu[i])/si[i])
print(a)
A=np.transpose(a)
#variance covariance matrix
VarCov=np.dot(np.transpose(A),A)/n1
print('\nVariance Covariance Matrix is :\n {}'.format(VarCov))
#calculating eigen values and vectors
eigenValues, eigenVectors = np.linalg.eig(VarCov)
idx = eigenValues.argsort()[::-1]
eigenValues = eigenValues[idx]
eigenVectors = eigenVectors[:,idx]
print('\nEigen Values are {}'.format(eigenValues))
print('\nEigen Vectors are {}'.format(eigenVectors))
```

```
#no of principal components to be retained
k=int(input("enter threshold limit :"))
s=sum(eigenValues)
t=[]
stoppoint=1
for i in range(len(eigenValues)):
    numerator=sum(eigenValues[0:i+1])
    z = (numerator/s) *100
    if z \le k:
        stoppoint=stoppoint+1
    t.append(z)
print('\nThreshold Table :\n{}'.format(t))
eigenValues=eigenValues[0:stoppoint]
eigenVectors=eigenVectors[:,0:stoppoint]
#retaining eigen values and vectors
print('\nRetained Eigen Values :\n{}'.format(eigenValues))
print('\nRetained Eigen Vectors: \n{}'.format(eigenVectors))
egv=np.transpose(eigenVectors)
f=[]
for i in range(len(egv)):
  0 = []
  for j in range(n):
    k=sqrt(eigenValues[i]) *eqv[i][j]
    o.append(k)
  f.append(o)
print(f)
h=[]
for i in range(len(f[0])):
  h.append(f[0][i]*2+f[1][i]*2)
print(h)
print("Variance:")
sumh=sum(h)
print(sumh)
pve=[]
for i in range(len(eigenValues)):
  pve.append((eigenValues[i]/sumh)*100)
print(pve)
```

Output 1:-

3 7 10 3 10 6 3 9 9 6 5 3 8 7 5

[6.6, 6.6, 5.6]

enter no of variables3

[3.5071, 2.51, 1.9494]

[[3, 7, 10, 3, 10], [6, 3, 9, 9, 6], [5, 3, 8, 7, 5]]

```
[array([-1.02648912, 0.11405435, 0.96946195, -1.02648912, 0.96946195]), array([-
0.23904382, -1.43426295, 0.9561753, 0.9561753, -0.23904382]), array([-0.30778701, -
1.33374372, 1.23114805, 0.71816969, -0.30778701])]
Variance Covariance Matrix is:
[[ 0.80001623 -0.04089598  0.06435815]
[-0.04089598 0.7999873 0.78479557]
[ 0.06435815  0.78479557  0.79996624]]
Eigen Values are [1.58512447 0.80666111 0.0081842 ]
Eigen Vectors are [[ 0.02122114 -0.99538368 -0.09360014]
[ 0.70624057  0.0811911  -0.70330098]
[ 0.70765382 -0.05117936  0.7047033 ]]
enter threshold limit:99
Threshold Table:
[66.04768471804098, 99.65898746868129, 100.0]
Retained Eigen Values:
[1.58512447 0.80666111]
Retained Eigen Vectors:
[[ 0.02122114 -0.99538368]
[ 0.70624057  0.0811911 ]
[ 0.70765382 -0.05117936]]
[[0.026717780654088277, 0.8891690658976457, 0.8909483734282817], [-
0.8939970315309378, 0.07292122996398519, -0.04596639553844041
[-1.734558501753699, 1.9241805917232617, 1.6899639557796826]
Variance:
1.8795860457492453
[84.33370077691795, 42.916955545008435]
```