다변량 통계 과제 2

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- 언어: python

```
import matplotlib.pyplot as plt
import numpy as np
import pandas as pd
import os
from glob import glob
import sys

import scipy.stats as stats # qqplot에 사용

In [2]: #시각화 패키지들
import seaborn as sns
import matplotlib.pyplot as plt
%matplotlib inline

plt.style.use("ggplot")
```

문제 1.

x1:석유, x2: 천연가스, x3:수력 발전, x4:원자력 발전 의 에너지 소비량의 표본 평균과 표본 공분산이다.

$$\bar{\mathbf{X}} = \begin{bmatrix} 0.766 \\ 0.508 \\ 0.438 \\ 0.161 \end{bmatrix}, \quad \mathbf{S} = \begin{bmatrix} 0.856 & 0.635 & 0.173 & 0.096 \\ 0.635 & 0.568 & 0.128 & 0.067 \\ 0.173 & 0.128 & 0.171 & 0.039 \\ 0.096 & 0.067 & 0.039 & 0.043 \end{bmatrix},$$

1 - a)

X1+X2+X3+X4의 표본평균과 표본분산

• E(X1+X2+X3+X4)=E(X1)+E(X2)+E(X3)+E(X4)

• VAR(X1+X2+X3+X4) = COV(X1+X2+X3+X4, X1+X2+X3+X4) = SUM(공분산 행렬) In [53]: $x \ all \ mean = x \ bar.T @ np.array([1,1,1,1])$ print("표본평균 : ", x all mean) 표본평균 : [1.873] In [54]: x all var=np.array([1,1,1,1]) @ S @ np.array([1,1,1,1])print("표본분산 : ",x all var) 표본분산 : 3.914 1 - b) X1-X2의 표본평균과 표본분산 • E(X1-X2) = E(X1)-E(X2) VAR(X1-X2) = VAR(X1)+VAR(X2)-2COV(X1,X2) In [55]: x diff mean = x bar.T @ np.array([1,-1,0,0])print("표본평균 : ", x diff mean) 표본평균: [0.258] In [56]: x diff var=np.array([1,-1,0,0]) @ S @ np.array([1,-1,0,0]) print("표본분산 : {:.4f}".format(x_diff_var)) 표본분산 : 0.1540 1 - c(X1+X2+X3+X4)과 (X1-X2)의 표본공분산(행렬) 구하기 COV(X1+X2+X3+X4,X1-X2) = {VAR(X1)+COV(X2,X1)+COV(X3,X1)+COV(X4,X1)} - $\{COV(X1,X2)+VAR(X2)+COV(X3,X2)+COV(X4,X2)\}$

• 공분산은 공분산행렬S에서 1열 - 2열 한것의 결과와 같다,

문제 2.

아래의 평균과 공분산을 갖는 이변량 정규분포의 n=1000 랜덤 표본 추출

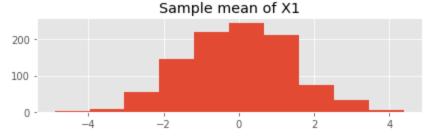
$$\mu = \begin{bmatrix} 0 \\ 0 \end{bmatrix}, \quad \Sigma = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}$$

x2<mark>의 평균: 0.0008866986650858212</mark>

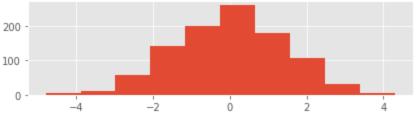
2 - a).

n=1000, rep=1000 xbar 와 S를 구해서 히스토그램과 수치로 비교

```
In [113...
         bivariate norm mean= np.array([0,0])
         bivariate norm cov= np.array([[2,1],[1,2]])
         bivariate norm cov
         array([[2, 1],
Out[113...
                [1, 2]])
In [142...
         \verb"np.stack([np.cov(x1,x2),np.cov(x1,x2),np.cov(x1,x2),np.cov(x1,x2)])[:,0,0]
         array([2.36427718, 2.36427718, 2.36427718, 2.36427718])
Out[142...
In [144...
         x1 means=[]
         x2 means=[]
         all cov=[]
         for nums in range(1000):
              x1,x2=np.random.multivariate normal(bivariate norm mean,bivariate norm cov,1000).T
              x1 means.append(np.mean(x1))
              x2 means.append(np.mean(x2))
              all cov.append(np.cov(x1,x2))
         all cov=np.stack(all cov)
In [145...
         print('x1의 평균 : ',np.mean(x1 means))
         print('x2의 평균 : ',np.mean(x2 means))
         plt.subplot(2,1,1)
         plt.title("Sample mean of X1")
         plt.hist(x1)
         plt.subplot(2,1,2)
         plt.title("Sample mean of X2")
         plt.hist(x2)
         plt.tight layout()
         x1의 평균: 0.002004897339473411
```

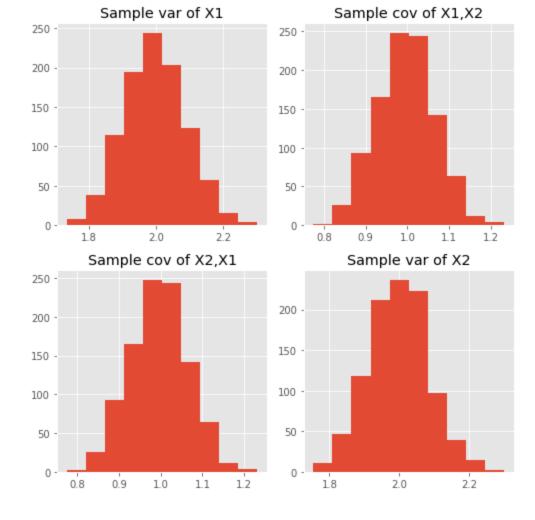


Sample mean of X2



```
In [154...
         print("S11 : ",np.mean(all cov[:,0,0]))
         print("S12 : ", np.mean(all cov[:,0,1]))
         print("S21 : ",np.mean(all cov[:,1,0]))
         print("S22 : ",np.mean(all cov[:,1,1]))
         plt.figure(figsize=(7,7))
         plt.subplot(2,2,1)
         plt.title("Sample var of X1")
         plt.hist(all cov[:,0,0])
         plt.subplot(2,2,2)
         plt.title("Sample cov of X1,X2")
         plt.hist(all cov[:,0,1])
         plt.subplot(2,2,3)
         plt.title("Sample cov of X2,X1")
         plt.hist(all cov[:,1,0])
         plt.subplot(2,2,4)
         plt.title("Sample var of X2")
         plt.hist(all cov[:,1,1])
         plt.tight layout()
```

S11 : 1.996653232770947
S12 : 0.9970238842444674
S21 : 0.9970238842444674
S22 : 1.9986437551327636

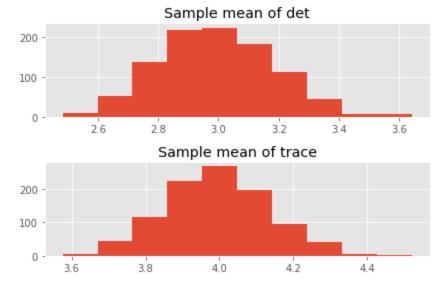


2 - b)

n=1000 표본에서 1000개의 |S| 와 tr(S)의 히스토그램 그리고, 평균을 나타내기

```
In [162...
          all cov.shape
         (1000, 2, 2)
Out[162...
In [174...
          all cov det=np.linalg.det(all cov)
          all cov trace=np.trace(all cov,axis1=1,axis2=2)
In [177...
          print("Mean of sample det : ", np.mean(all cov det))
          print("Mean of sample trace : ",np.mean(all cov trace))
          plt.subplot(2,1,1)
          plt.title("Sample mean of det")
          plt.hist(all_cov_det)
          plt.subplot(2,1,2)
          plt.title("Sample mean of trace")
          plt.hist(all_cov_trace)
          plt.tight layout()
```

Mean of sample det : 2.993990140552404
Mean of sample trace : 3.99529698790371



실제 determinant : 3, trace : 4로, 거의 일치한 수치를 얻을 수 있다.

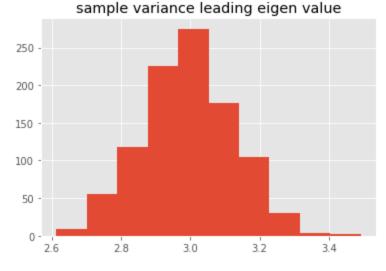
2 - c)

1000개 표본, 1000 rep 에서 S의 leading eigen value 구하고 비교

```
print("실제 cov의 eigen value : ",np.linalg.eigvals(bivariate_norm_cov))
실제 cov의 eigen value : [3. 1.]

In [209... all_cov_leading_eigval =np.sort(np.linalg.eigvals(all_cov),)[:,1]
    print("Mean of sample variance leading eigen value",np.mean(all_cov_leading_eigval))
    plt.title("sample variance leading eigen value")
    plt.hist(all_cov_leading_eigval)
    plt.show()
```

Mean of sample variance leading eigen value 2.9960892241327493



실제값 3과 매우 유사함을 알 수 있다.

문제 3.

• T4-6.DAT 130명 페루 청소년 심리검사 결과

• 첫 5개 변수 : independence, support, benevolence, conformity, leadership

Out[224... independence support benevolence conformity leadership 0 27.0 13.0 14.0 20.0 11.0 1 12.0 25.0 6.0 13.0 24.0 2 14.0 20.0 15.0 16.0 7.0 3 18.0 20.0 17.0 12.0 6.0 9.0 22.0 22.0 4 21.0 6.0 ••• 125 10.0 11.0 26.0 17.0 10.0 126 14.0 12.0 14.0 29.0 11.0 127 19.0 11.0 23.0 18.0 13.0 128 27.0 19.0 22.0 7.0 9.0

17.0

22.0

130 rows × 5 columns

3 - a).

129

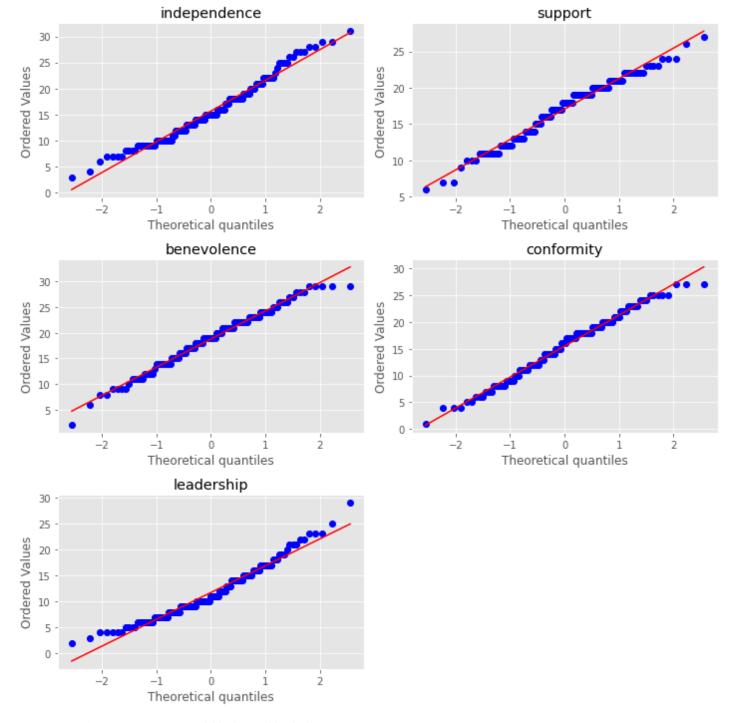
Q-Q PLOT marginal normality

10.0

```
In [235...
         plt.figure(figsize=(10,10))
         plt.subplot(3,2,1)
         stats.probplot(X['independence'],plot=plt)
         plt.title('independence')
         plt.subplot(3,2,2)
         stats.probplot(X['support'],plot=plt)
         plt.title('support')
         plt.subplot(3,2,3)
         stats.probplot(X['benevolence'],plot=plt)
         plt.title('benevolence')
         plt.subplot(3,2,4)
         stats.probplot(X['conformity'],plot=plt)
         plt.title('conformity')
         plt.subplot(3,2,5)
         stats.probplot(X['leadership'],plot=plt)
         plt.title('leadership')
         plt.tight layout()
```

22.0

8.0



Indepence와 Leadership은 곡선형태로 관측된다.

Indepence와 Leadership은 정규성을 띄지않는다고 생각된다.

3 - b).

• Chi-square PLOT multivariate noramlity

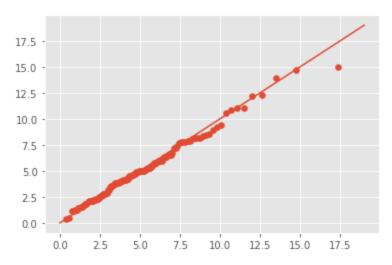
```
deviation_matrix=X-X.mean()
    cov_matrix = deviation_matrix.cov().to_numpy()
    invcov_matrix=np.linalg.inv(cov_matrix)

chisq_list =[]

for i in range(130):
    row_one=deviation_matrix.to_numpy()[i,:]
    chisq_list.append(row_one.T @ invcov_matrix @ row_one)
```

Out[294...

[<matplotlib.lines.Line2D at 0x1911ea61488>]



multivariate Chi-square plot을 그려본 결과.

마지막 한 값이 아웃라이어가 존재한다.

그것을 제외하면, 정규성을 띄는것으로 관측된다.