

I. 포아송분포

II. 이항분포

# I. 포아송분포

## 1. 확률함수

- 단위구간 내에서 어떤 사건이 평균  $\mu$  회 발생하고, 확률변수  $X$ 를 사건의 발생횟수라고 할 때  
 $X \sim \text{Poisson}(\mu)$
- 확률질량함수는 다음과 같음  
$$P_r(x=k) = \frac{\mu^k}{k!} e^{-\mu}, \quad 0, 1, 2, \dots$$
- 포아송분포의 모수는 평균( $\mu$ )
- 포아송분포의 모양은 평균이 작을 때는  
좌우비대칭이나  
평균이 증가함에 따라 평균을 중심으로  
좌우대칭의 모양으로 변함.  
즉, 정규분포에 가까워지고, 분산은 커짐

b1-ch4-5-r.py

```
import numpy as np
#import scipy.stats as stats
import seaborn as sns # seaborn package를 이용
import matplotlib.pyplot as plt # matplotlib.pyplot package를 이용
from numpy import random

# set the random seed:
np.random.seed(12345)
r=10000

poi1 = random.poisson(3, size=r)
sns.histplot(data=poi1, x=None).set(title='Histogram of Poi3')
plt.show()

poi2 = random.poisson(5.0, size=r)
sns.histplot(data=poi2, x=None).set(title='Histogram of Poi5')
plt.show()

poi3 = random.poisson(10.0, size=r)
sns.histplot(data=poi3, x=None).set(title='Histogram of Poi10')
plt.show()

poi4 = random.poisson(16, size=r)
sns.histplot(data=poi4, x=None).set(title='Histogram of Poi16')
plt.show()
```

## b1-ch4-5-r.py

(앞에서 계속)

```
fig = plt.figure(figsize=(14,7))

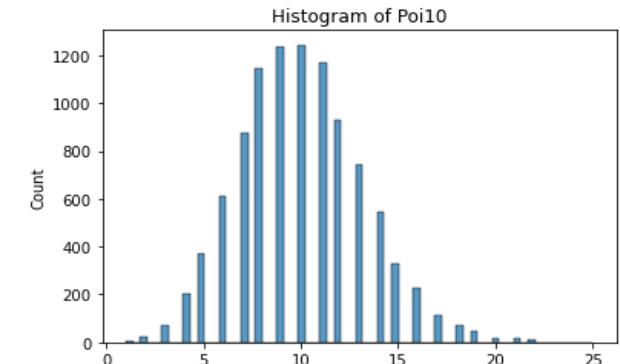
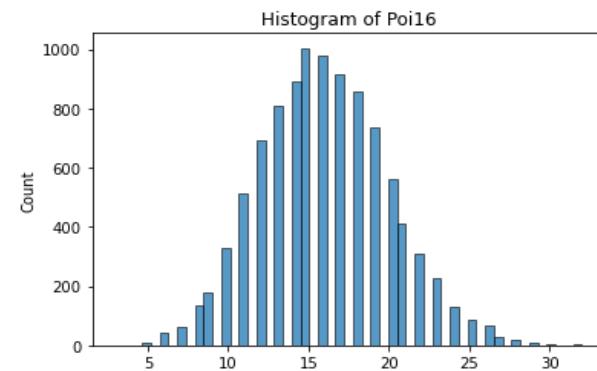
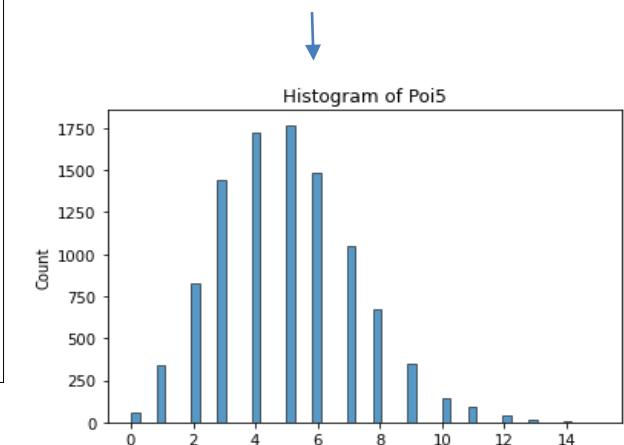
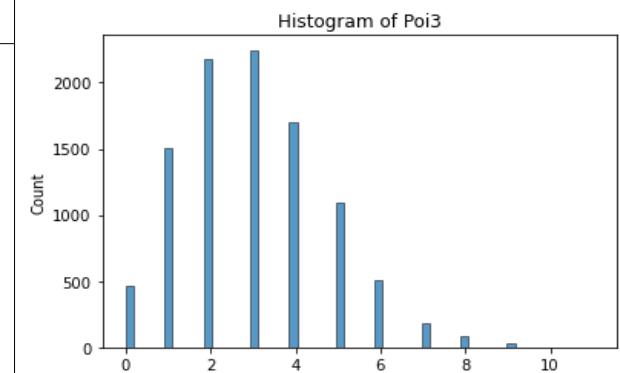
# 두 개의 그래프를 한 페이지에 그림
fig, axs = plt.subplots(ncols=2)
fig, ays = plt.subplots(ncols=2)

sns.histplot(data=poi1, x=None, ax=axs[0]).set(title='Histogram of Poisson(mu=3.0 & mu=5.0)')
sns.histplot(data=poi2, x=None, ax=axs[1])

fig.subplots_adjust(wspace=0.5) # 우측그림의 좌우 간격을 조정
plt.savefig('C:/BOOK/PyBasics/PyStat/code/poison-1.png')

sns.histplot(data=poi3, x=None, ax=ays[0]).set(title='Histogram of Poisson(mu=10.0 & mu=16.0)')
sns.histplot(data=poi4, x=None, ax=ays[1])

fig.subplots_adjust(wspace=0.5) # 우측그림의 좌우 간격을 조정
plt.savefig('C:/BOOK/PyBasics/PyStat/code/poison-2.png')
```



## b1-ch4-6.py

```
import scipy.stats as stats
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
# values for x (all between 0 and 30):
x = np.linspace(0, 30, num=31)
# PMF for all these values:
fx_3 = stats.poisson.pmf(x, 3)
fx_5 = stats.poisson.pmf(x, 5)
fx_10 = stats.poisson.pmf(x, 10)
fx_15 = stats.poisson.pmf(x, 15)
# collect values in DataFrame:
result_3 = pd.DataFrame({'x': x, 'fx_3': fx_3})
result_5 = pd.DataFrame({'x': x, 'fx_5': fx_5})
result_10 = pd.DataFrame({'x': x, 'fx_10': fx_10})
result_15 = pd.DataFrame({'x': x, 'fx_15': fx_15})
#print(f'result_3: \n{result_3}\n')
#print(f'result_5: \n{result_5}\n')
#print(f'result_10: \n{result_10}\n')
#print(f'result_15: \n{result_15}\n')
"""
# plot:
plt.bar(x, fx_3, color='0.6', width = 0.1)
plt.xlabel('x')
plt.ylabel('P(X=x)')
plt.show()
```

## b1-ch4-6.py

(앞에서 계속)

```
Plt.bar(x, fx_5, color= ' 0.6 ' , width = 0.1)
Plt.|뮤디( ' x ' )
Plt.ylabel( ' P(X=x) ' )
Plt.show()
plt.bar(x, fx_10, color='0.6', width = 0.1)
plt.xlabel('x')
plt.ylabel('P(X=x)')
plt.show()
plt.bar(x, fx_15, color='0.6', width = 0.1)
plt.xlabel('x')
plt.ylabel('P(X=x)')
plt.show()
"""

# create graph:
fig = plt.figure(figsize=(14,7))
plt.subplot(221).set(title='Probability of Poisson with lambda=3')
_= plt.bar(x, fx_3, color='0.6', width = 0.1)
#_= plt.stem(x, fx_3, use_line_collection = True)
_= plt.xlabel('x')
_= plt.ylabel('P(X=x)')
plt.subplot(222).set(title='Probability of Poisson with lambda=5')
_= plt.bar(x, fx_5, color='0.6', width = 0.1)
#_= plt.stem(x, fx_5, use_line_collection = True)
_= plt.xlabel('x')
_= plt.ylabel('P(X=x)')
```

## b1-ch4-6.py

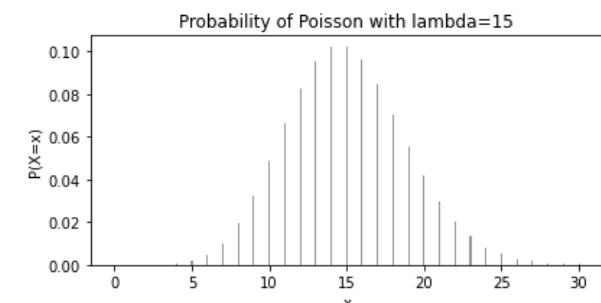
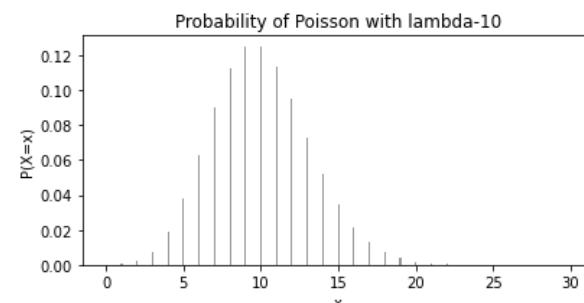
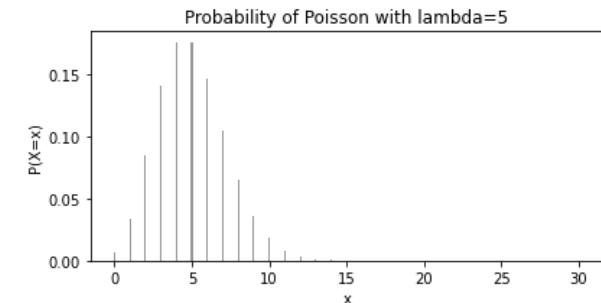
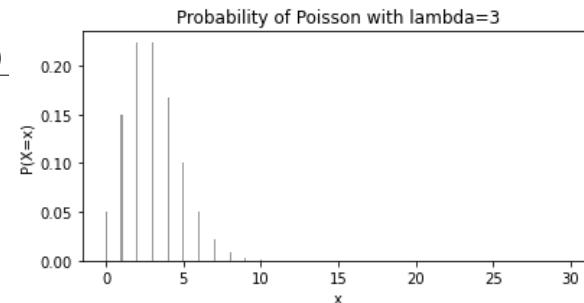
(앞에서 계속)

```
plt.subplot(223).set(title='Probability of Poisson with lambda=10')
_ = plt.bar(x, fx_10, color='0.6', width = 0.1)
#_ = plt.stem(x, fx_10, use_line_collection = True)
_ = plt.xlabel('x')
_ = plt.ylabel('P(X=x)')

plt.subplot(224).set(title='Probability of Poisson with lambda=15')
_ = plt.bar(x, fx_15, color='0.6', width = 0.1)
#_ = plt.stem(x, fx_15, use_line_collection = True)
_ = plt.xlabel('x')
_ = plt.ylabel('P(X=x)')

fig.subplots_adjust(hspace=0.4, wspace=0.2) # 하측그림의 상하
# 및 우측그림의 좌우 간격을 조정

plt.show()
# plt.savefig('D:/BOOK/PyBasics/PyStat/code/b1-ch4-6.png')
```



## 2. 확률분포표

b1-ch4-7.py

```
import pandas as pd
import numpy as np
from scipy.stats import poisson
r = 5
poisson_02 = np.empty(r)
poisson_04 = np.empty(r)
poisson_06 = np.empty(r)
poisson_08 = np.empty(r)
poisson_1 = np.empty(r)
poisson_2 = np.empty(r)
poisson_3 = np.empty(r)
poisson_4 = np.empty(r)
# repeat r times:
for j in range(r):
    poisson_02[j] = poisson.cdf(k=j, mu=0.02)
    j=j+1
for j in range(r):
    poisson_04[j] = poisson.cdf(k=j, mu=0.04)
    j=j+1
for j in range(r):
    poisson_06[j] = poisson.cdf(k=j, mu=0.06)
    j=j+1
for j in range(r):
    poisson_08[j] = poisson.cdf(k=j, mu=0.08)
    j=j+1
```

## 2. 확률분포표

b1-ch4-7.py

(앞에서 계속)

```
for j in range(r):
    poisson_1[j] = poisson.cdf(k=j, mu=0.1)
```

j=j+1

```
for j in range(r):
```

```
    poisson_2[j] = poisson.cdf(k=j, mu=0.2)
```

j=j+1

```
for j in range(r):
```

```
    poisson_3[j] = poisson.cdf(k=j, mu=0.3)
```

j=j+1

```
for j in range(r):
```

```
    poisson_4[j] = poisson.cdf(k=j, mu=0.4)
```

j=j+1

```
poisson_mu = pd.DataFrame({'0.02':poisson_02,'0.04':poisson_04, '0.06':poisson_06,'0.08':poisson_08, '0.1':poisson_1,'0.2':poisson_2,'0.3':poisson_3,'0.4':poisson_4})
```

```
round(poison_mu, 3)
```

```
print("Probability Distribution Table of Poisson with 'mu=0.02, ..., 0.4' : ", f'\n{round(poison_mu, 3)}\n')
```

c \ $\mu$	0.02	0.04	0.06	0.08	0.10	0.20	0.30	0.40
0	0.980	0.961	0.942	0.923	0.905	0.819	0.741	0.670
1	1.000	0.999	0.998	0.997	0.995	0.982	0.963	0.938
2	1.000	1.000	1.000	1.000	1.000	0.999	0.996	0.992
3	1.000	1.000	1.000	1.000	1.000	1.000	1.000	0.999
4	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000

```
Probability Distribution Table of Poisson with 'mu=0.02, ..., 0.4' :
 0.02  0.04  0.06  0.08  0.1   0.2   0.3   0.4
0  0.98  0.961 0.942 0.923 0.905 0.819 0.670 0.670
1  1.00  0.999 0.998 0.997 0.995 0.982 0.938 0.938
2  1.00  1.000 1.000 1.000 1.000 0.999 0.992 0.992
3  1.00  1.000 1.000 1.000 1.000 1.000 0.999 0.999
4  1.00  1.000 1.000 1.000 1.000 1.000 1.000 1.000
```

## II. 이항분포

### 1. 확률함수

- 성공의 확률이  $p$ 이고 실패의 확률이  $q(q = 1-p)$ 인 베르누이 시행을 독립적으로  $n$ 번 반복하였을 때 나타나는 성공의 횟수를 확률변수  $X$ 라고 할 때

$$X \sim B(n, p)$$

- 확률질량함수는 다음과 같음

$$P_r(X = k) = \binom{n}{k} P^k q^{n-k}, \quad k = 0, 1, 2, \dots, n$$

- 이항분포의 모수는 각 시행에서의 성공확률  $p$ 와 시행횟수  $n$
- 이항분포는  $n$ 과  $p$ 의 크기에 따라 모양이 결정되는데  $p$ 가 0.5이면 평균( $\mu = np = \frac{n}{2}$ )을 중심으로 좌우 대칭인 분포를 가지며,  $p=0.5$ 이고  $n$ 이 커질 때 좌우대칭의 정규분포와 유사함

b1-ch4-2.py

```
import numpy as np
import seaborn as sns # seaborn package를 이용
import matplotlib.pyplot as plt # matplotlib.pyplot package를 이용
from numpy import random

# set the random seed:
np.random.seed(12345)
r=10000

binom1 = random.binomial(n=10, p=0.1, size=r)
sns.histplot(data=binom1, x=None).set(title='Histogram of Binom1')
plt.show()

binom2 = random.binomial(n=10, p=0.5, size=r)
sns.histplot(data=binom2, x=None).set(title='Histogram of Binom2')
plt.show()

binom3 = random.binomial(n=20, p=0.5, size=r)
sns.histplot(data=binom3, x=None).set(title='Histogram of Binom3')
plt.show()
```

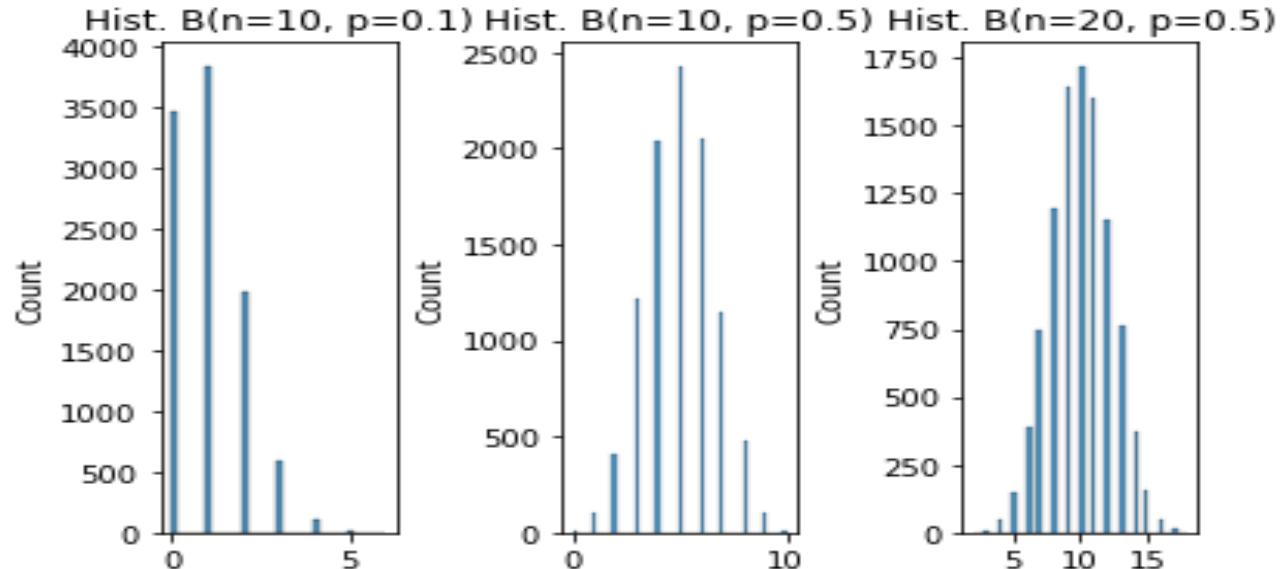
## b1-ch4-2.py

(앞에서 계속)

```
#fig = plt.figure(figsize=(10,7))
#figure, axs = plt.subplots(1,3) # 세 개의 그래프를 한 페이지에 그림
fig, axs = plt.subplots(ncols=3)

sns.histplot(data=binom1, x=None, ax=axs[0]).set(title='Hist. B(n=10, p=0.1)')
sns.histplot(data=binom2, x=None, ax=axs[1]).set(title='Hist. B(n=10, p=0.5)')
sns.histplot(data=binom3, x=None, ax=axs[2]).set(title='Hist. B(n=20, p=0.5)')

fig.subplots_adjust(wspace=0.7) # 우측그림의 좌우 간격을 조정
#plt.savefig('C:/BOOK/PyBasics/PyStat/code/binom.png')
```



## b1-ch4-3.py

```
import scipy.stats as stats
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
# values for x (all between 0 and 25):
x = np.linspace(0, 25, num=26)
# PMF for all these values:
fx_01 = stats.binom.pmf(x, 25, 0.1)
fx_02 = stats.binom.pmf(x, 25, 0.2)
fx_05 = stats.binom.pmf(x, 25, 0.5)
fx_07 = stats.binom.pmf(x, 25, 0.7)
# collect values in DataFrame:
result_01 = pd.DataFrame({'x': x, 'fx_01': fx_01})
result_02 = pd.DataFrame({'x': x, 'fx_02': fx_02})
result_05 = pd.DataFrame({'x': x, 'fx_05': fx_05})
result_07 = pd.DataFrame({'x': x, 'fx_07': fx_07})
"""
# plot:
plt.bar(x, fx_01, color='0.6', width = 0.1)
plt.xlabel('x')
plt.ylabel('P(X=x)')
plt.show()
plt.bar(x, fx_02, color='0.6', width = 0.1)
plt.xlabel('x')
plt.ylabel('P(X=x)')
plt.show()
```

## b1-ch4-3.py

(앞에서 계속)

```
Plt.bar(x, fx_05, color= ' 0.6 ' , width = 0.1)
```

```
Plt.xlabel( ' x ' )
```

```
Plt.ylabel( ' P(X=x) ' )
```

```
Plt.show()
```

```
Plt.bar(x, fx_07, color= ' 0.6 ' , width = 0.1)
```

```
Plt.xlabel( ' x ' )
```

```
Plt.ylabel( ' P(X=x) ' )
```

```
Plt.show()
```

```
" " "
```

```
# create graph:
```

```
Fig = plt.figure(figsize=(14,7))
```

```
plt.subplot(221).set(title='Probability of Binomial with n=25 and p=0.1')
```

```
_ = plt.bar(x, fx_01, color='0.6', width = 0.1)
```

```
#_ = plt.stem(x, fx_01, use_line_collection = True)
```

```
_ = plt.xlabel('x')
```

```
_ = plt.ylabel('P(X=x)')
```

```
plt.subplot(222).set(title='Probability of Binomial with n=25 and p=0.2')
```

```
_ = plt.bar(x, fx_02, color='0.6', width = 0.1)
```

```
#3_ = plt.stem(x, fx_02, use_line_collection = True)
```

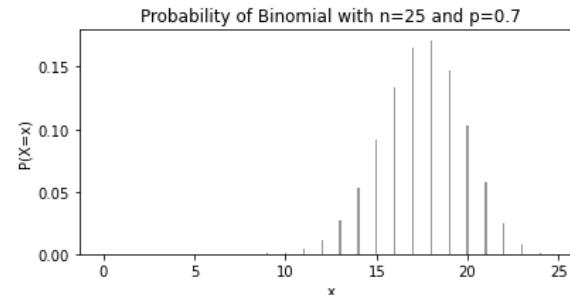
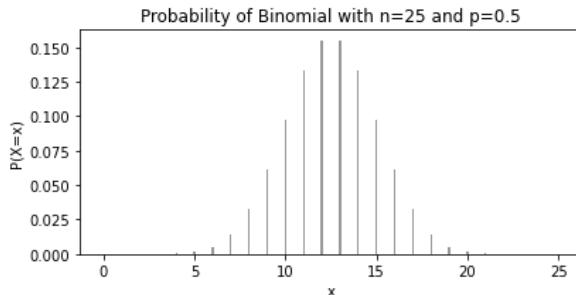
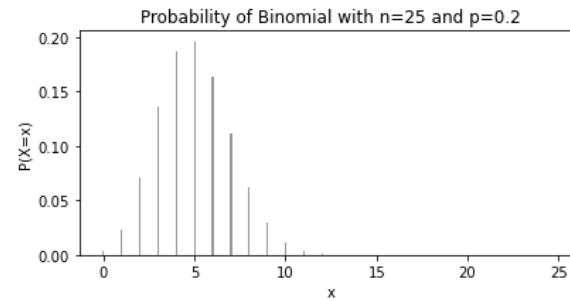
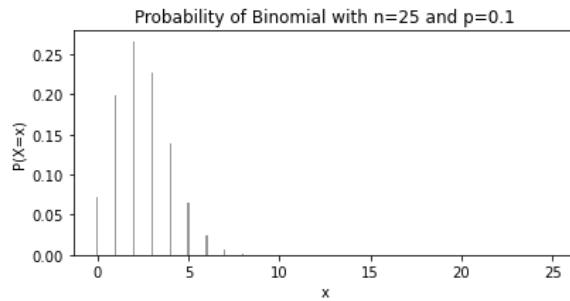
```
_ = plt.xlabel('x')
```

```
_ = plt.ylabel('P(X=x)')
```

## b1-ch4-3.py

(앞에서 계속)

```
plt.subplot(223).set(title='Probability of Binomial with n=25 and p=0.5')
_ = plt.bar(x, fx_05, color='0.6', width = 0.1)
#3_ = plt.stem(x, fx_05, use_line_collection = True)
_ = plt.xlabel('x')
_ = plt.ylabel('P(X=x)')
plt.subplot(224).set(title='Probability of Binomial with n=25 and p=0.7')
_ = plt.bar(x, fx_07, color='0.6', width = 0.1)
#_ = plt.stem(x, fx_07, use_line_collection = True)
_ = plt.xlabel('x')
_ = plt.ylabel('P(X=x)')
fig.subplots_adjust(hspace=0.4, wspace=0.2) # 하측그림의 상하 및 우측그림의 좌우 간격을 조정
#plt.savefig('C:/BOOK/PyBasics/PyStat/code/binom_pmf.png')
```



## 2. 확률분포표

b1-ch4-4.py

```
import pandas as pd
import numpy as np
#import scipy.stats as stats
from scipy.stats import binom
r = 5
binom_01 = np.empty(r)
binom_02 = np.empty(r)
binom_03 = np.empty(r)
binom_04 = np.empty(r)
binom_05 = np.empty(r)
binom_06 = np.empty(r)
binom_07 = np.empty(r)
binom_08 = np.empty(r)
binom_09 = np.empty(r)
# repeat r times:
for j in range(r):
    binom_01[j] = binom.cdf(k=j, n=5, p=0.1)
    j=j+1
for j in range(r):
    binom_02[j] = binom.cdf(k=j, n=5, p=0.2)
    j=j+1
for j in range(r):
    binom_03[j] = binom.cdf(k=j, n=5, p=0.3)
    j=j+1
```

## b1-ch4-4.py

(앞에서 계속)

For j in range<sup>®</sup>:

    binom\_04[j] = binom.cdf(k=j, n=5, p=0.4)

    j=j+1

for j in range(r):

    binom\_05[j] = binom.cdf(k=j, n=5, p=0.5)

    j=j+1

for j in range(r):

    binom\_05[j] = binom.cdf(k=j, n=5, p=0.6)

    j=j+1

for j in range(r):

    binom\_07[j] = binom.cdf(k=j, n=5, p=0.7)

    j=j+1

for j in range(r):

    binom\_08[j] = binom.cdf(k=j, n=5, p=0.8)

    j=j+1

for j in range(r):

    binom\_09[j] = binom.cdf(k=j, n=5, p=0.9)

    j=j+1

for j in range(r):

    binom\_095[j] = binom.cdf(k=j, n=5, p=0.95)

    j=j+1

binom\_n5 = pd.DataFrame({'0.10':binom\_01,'0.20':binom\_02,'0.30':binom\_03,'0.40':binom\_04,'0.50':binom\_05,'0.60':binom\_07,'0.70':binom\_07,'0.80':binom\_08,'0.90':binom\_09,'0.95':binom\_095 })

round(binom\_n5,3)

print("Probability Distribution Table of Binomial with 'n=5 and p=0.10, ..., 0.95' : ", f'\n{round(binom\_n5,3)}\n')

n=5

a	0.10	0.20	0.30	0.40	0.50	0.60	0.70	0.80	0.90	0.95
0	.590	.328	.168	.078	.031	.010	.002	.000	.000	.000
1	.919	.737	.528	.337	.187	.087	.031	.007	.000	.000
2	.991	.942	.837	.683	.500	.317	.163	.058	.009	.001
3	1.00	.993	.969	.913	.813	.663	.472	.263	.081	.023
4	1.00	1.00	.998	.990	.989	.922	.832	.672	.410	.226

Probability Distribution Table of Binomial with 'n=5 and p=0.10, ..., 0.95' :

0.10    0.20    0.30    0.40    0.50    0.60    0.70    0.80    0.90    0.95

0    0.590    0.328    0.168    0.078    0.031    0.010    0.002    0.000    0.000    0.000

1    0.919    0.737    0.528    0.337    0.187    0.087    0.031    0.007    0.000    0.000

2    0.991    0.942    0.837    0.683    0.500    0.317    0.163    0.058    0.009    0.001

3    1.000    0.993    0.969    0.913    0.813    0.663    0.472    0.263    0.081    0.023

4    1.000    1.000    0.998    0.990    0.989    0.922    0.832    0.672    0.410    0.226