Assignment Binomial Distribution

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1. For a Binomial Distribution parameter n=5 and p=0.3 Find the probabilities of getting

a) At least 3 successes

The probability that a random variable X with binomial distribution B(n, p) is equal to the value k, where k = 0, 1, ... n, is given by

$$P(X = k) = \binom{n}{k} p^k (1 - p)^{n-k} = \frac{n!}{k!(n-k)!} p^k (1 - p)^{n-k}$$

$$P(X \ge 3) = P(X = 3) + P(X = 4) + P(X = 5)$$

$$= {5 \choose 3} (0.3)^3 (1 - 0.3)^2 + {5 \choose 4} (0.3)^4 (1 - 0.3) + {5 \choose 5} (0.3)^5$$

$$= {5! \over 3!(5 - 3)!} (0.3)^3 (0.7)^2 + {5! \over 4!(5 - 4)!} (0.3)^4 (0.7) + {5! \over 5!(5 - 5)!} (0.3)^5$$

$$= 0.1323 + 0.0283 + 0.0024$$

$$= 0.1630$$

b) At most 3 successes

Method 1:

$$P(X \le 3) = P(X = 0) + P(X = 1) + P(X = 2) + P(X = 3)$$

$$= {5 \choose 0} (1 - 0.3)^5 + {5 \choose 1} (0.3) (1 - 0.3)^4 + {5 \choose 2} (0.3)^2 (1 - 0.3)^3 + {5 \choose 3} (0.3)^3 (1 - 0.3)^2$$

$$= {5! \over 0!(5 - 0)!} (0.7)^5 + {5! \over 1!(5 - 1)!} (0.3) (0.7)^4 + {5! \over 2!(5 - 2)!} (0.3)^2 (0.7)^3 + {5! \over 3!(5 - 3)!} (0.3)^3 (0.7)^2$$

$$= 0.1681 + 0.3601 + 0.3087 + 0.1323$$

$$= 0.9692$$

Method 2:

$$P(X \le 3) = 1 - P(X > 3)$$

$$= 1 - (P(X = 4) + P(X = 5))$$

$$= 1 - \left(\binom{5}{4}(0.3)^4(1 - 0.3) + \binom{5}{5}(0.3)^5\right)$$

$$= 1 - \left(\frac{5!}{4!(5 - 4)!}(0.3)^4(0.7) + \frac{5!}{5!(5 - 5)!}(0.3)^5\right)$$

$$= 1 - (0.0283 + 0.0024)$$

$$= 0.9692$$

c) Exactly 3 failures

$$P(X = 2) = {5 \choose 2} (0.3)^2 (1 - 0.3)^3$$
$$= \frac{5!}{2!(5-2)!} (0.3)^2 (0.7)^3$$
$$= 0.3087$$

Program:

```
#!/usr/bin/env python3
def fact(n):
    fact = 1
    for value in range(1, n+1):
        fact *= value
    return(fact)
              _____Binomial Distribution_____\n")
n = int(input("Number of trials: "))
p = float(input("Probability of getting a success in one trial: "))
option1 = input("Type L for at least, M for at most, E for exact: ")
option2 = input("Type S for success, F for failure: ")
if option2 == 'S':
    x = int(input("Number of success desired: "))
    chosen = 'success'
    k = x
else:
    x = int(input("Number of failure desired: "))
    chosen = 'failure'
    k = n-x
if option1 == 'L':
    sum = 0
    while (n >= k):
        sum += fact(n)/(fact(n-k)*fact(k)) * (p)**(k) * (1-p)**(n-k)
   print("\nThe probability of getting at least {} {}: {:.4f}\n".format(x, chosen, sum))
if option1 == 'M':
    sum = 0
    while (k >= 0):
        sum += fact(n) / (fact(n-k) * fact(k)) * (p) * * (k) * (1-p) * * (n-k)
    print("\nThe probability of getting at most {} {}: {:.4f}\n".format(x, chosen, sum))
if option1 == 'E':
    sum = fact(n) / (fact(n-k) * fact(k)) * (p) * * (k) * (1-p) * * (n-k)
   print("\nThe probability of getting exactly {} {}: {:.4f}\n".format(x, chosen, sum))
```

```
Output:
```

```
__Binomial Distribution___
Number of trials: 5
Probability of getting a success in one trial: 0.3
Type L for at least, M for at most, E for exact: L
Type S for success, F for failure: S
Number of success desired: 3
The probability of getting at least 3 success: 0.1631
        ___Binomial Distribution___
Number of trials: 5
Probability of getting a success in one trial: 0.3
Type L for at least, M for at most, E for exact: {\tt M}
Type S for success, F for failure: S
Number of success desired: 3
The probability of getting at most 3 success: 0.9692
         __Binomial Distribution___
Number of trials: 5
Probability of getting a success in one trial: 0.3
Type L for at least, M for at most, E for exact: {\tt E}
Type S for success, F for failure: F
Number of failure desired: 3
The probability of getting exactly 3 failure: 0.3087
Program: Binomial Distribution using Scipy package
#!/usr/bin/env python3
from scipy.stats import binom
# compute survival function (SF)
\# P(X >= 3)
pls = binom.sf(k=3, n=5, p=0.3, loc=1)
print("The probability of getting at least 3 success: {:.4f}".format(pls))
# compute cumulative density function (CDF)
# P(X \le 3)
pms = binom.cdf(k=3, n=5, p=0.3, loc=0)
print("The probability of getting at most 3 success: {:.4f}".format(pms))
# compute probability mass function (PMF)
\# P(X = 2)
pef = binom.pmf(k=2, n=5, p=0.3, loc=0)
print("The probability of getting exactly 3 failure: {:.4f}".format(pef))
Output:
The probability of getting at least 3 success: 0.1631
The probability of getting at most 3 success: 0.9692
The probability of getting exactly 3 failure: 0.3087
```

2. If on an average one vessel in every ten is wrecked, find the probability that out of five vessels expected to arrive, four at least will arrive safely.

The probability of success, p = 0.9 and the number of observations, n = 5.

$$P(X \ge 4) = P(X = 4) + P(X = 5)$$

$$= {5 \choose 4} (0.9)^4 (1 - 0.9) + {5 \choose 5} (0.9)^5$$

$$= {5! \over 4!(5 - 4)!} (0.9)^4 (0.1) + {5! \over 5!(5 - 5)!} (0.9)^5$$

$$= 0.3280 + 0.5905$$

$$= 0.9185$$

Program:

```
#!/usr/bin/env python3
from scipy.stats import binom

# probability of success, p = 0.9 and number of observations, n = 5
# P(X >= 4)
pls = binom.sf(k=4, n=5, p=0.9, loc=1)
print("The probability of getting at least 4 success is {:.4f}".format(pls))
Output:
```

The probability of getting at least 4 success is 0.9185

3. Five coins are tossed 3,200 times.

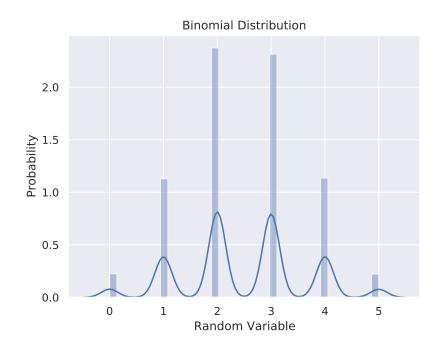
Program:

```
#!/usr/bin/env python3
import numpy as np
import pandas as pd
from scipy.stats import binom
import matplotlib.pyplot as plt
import seaborn as sns
sns.set()
data = binom.rvs(n=5, p=0.5, size=3200)
heads, freq = np.unique(data, return_counts=True)
dframe = pd.DataFrame({'Heads':heads, 'Frequencies':freq})
print("Frequencies of distribution of heads: \n{}".format(dframe))
print("Mean: {:.3f}".format(data.mean()))
print("Standard Deviation: {:.3f}".format(data.std()))
plt.figure(dpi=120)
sns.distplot(data)
plt.xlabel("Random Variable")
plt.ylabel("Probability")
plt.title("Binomial Distribution")
plt.savefig("binormdistplot.pdf")
plt.show()
```

a) Find the Frequencies of the distribution of heads and tabulate the results.

Frequencies of distribution of heads:

eads	Frequencies
0	97
1	488
2	1027
3	1001
4	491
5	96



b) Calculate the mean number of sucess and standard deviation.

Mean,
$$\mu_X=np=5\times0.5=2.5$$

Variance, $\sigma_X^2=np(1-p)=5\times0.5\times(1-0.5)=1.25$
Standard Deviation, $\sigma_X=\sqrt{np(1-p)}=\sqrt{1.25}=1.12$

Output:

Mean: 2.496

Standard Deviation: 1.106