1

Compulsory Assignment-Probability and Random Variable

Annu-EE21RESCH01010

Download latex code from here-

https://github.com/annu100/AI5002-Probabilityand-Random-variables/tree/main.tex/ Compulsory Assignment

Download python code from here-

https://github.com/annu100/AI5002-Probabilityand-Random-variables/tree/main.py/ Compulsory Assignment

I. Problem Statement-Problem 3.7

Let X represent the difference between the number of heads and the number of tails obtained when a coin is tossed 6 times. What are possible values of X?

II. SOLUTION

Let X_1 denotes the number of heads and X_2 denotes the number of tails that occur when a coin is tossed 6 times.

let n is total number of tosses and p is probability of getting head

$$p=q=0.5$$

Clearly, $X_1 \sim Bin(n = 6, p)$

and $X_2 \sim Bin(n = 6, 1 - p = q)$.

 $\therefore n - X_2 \sim Bin(6, p).$

By reproductive property,

$$X_1 + n - X_2 \sim Bin(6 + 6, p)$$
 (1)

 $X = X_1 - X_2.$

$$P(X = x) = \binom{12}{6+x} \frac{1}{2}^{12}, x = -6 \text{ to } 6$$

therefore,X can have any values between -6 to 6.

Convolution of Bernoulli distributions

The convolution of two independent identically distributed Bernoulli random variables is a binomial random variable.

$$\sum_{i=1}^{2} \text{Bernoulli}(p) \sim \text{Binomial}(2, p)$$

To show this let

 $X_i \sim \text{Bernoulli}(p), \quad 0$

and define

$$Y = \sum_{i=1}^{2} X_i$$

Also, let Z denote a generic binomial random variable:

 $Z \sim \text{Binomial}(2, p)$

Using probability mass functions

As X_1 and X_2 are independent,

$$\mathbb{P}[Y = n] = \mathbb{P}\left[\sum_{i=1}^{2} X_{i} = n\right]
= \sum_{m \in \mathbb{Z}} \mathbb{P}[X_{1} = m] \times \mathbb{P}[X_{2} = n - m]
= \sum_{m \in \mathbb{Z}} \left[\binom{1}{m} p^{m} (1 - p)^{1 - m}\right] \left[\binom{1}{n - m} p^{n - m} (1 - p)^{1 - n + n}\right]
= p^{n} (1 - p)^{2 - n} \sum_{m \in \mathbb{Z}} \binom{1}{m} \binom{1}{n - m}
= p^{n} (1 - p)^{2 - n} \left[\binom{1}{0} \binom{1}{n} + \binom{1}{1} \binom{1}{n - 1}\right]
= \binom{2}{n} p^{n} (1 - p)^{2 - n} = \mathbb{P}[Z = n]$$

Thus, convolution of 2 binomial distributions is also binomial

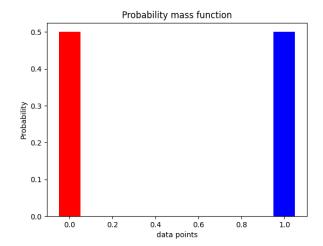


Figure 1: Probability mass function

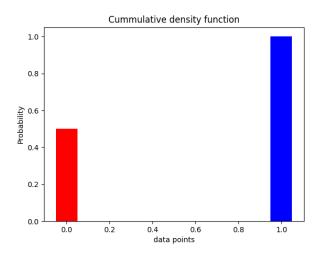
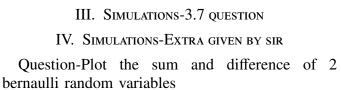
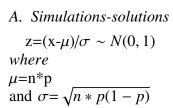


Figure 2: Cummulative mass function





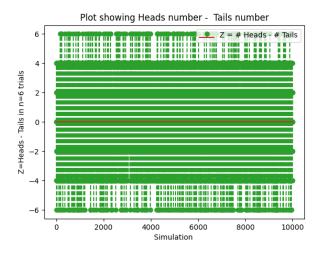


Figure 3: Number of heads -numbrer of tails plot

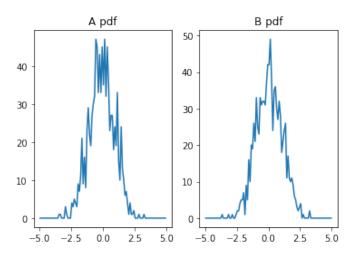


Figure 4: Cummulative mass function

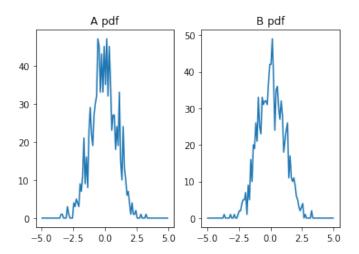


Figure 5: pdf A and B

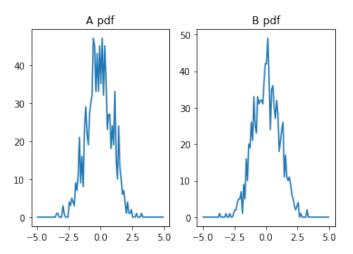


Figure 6: pdf of A and B mass function

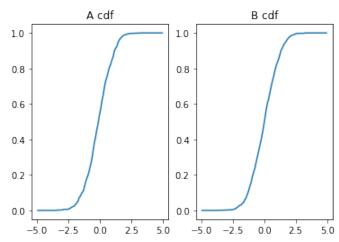


Figure 9: Cummulative mass function of A and B

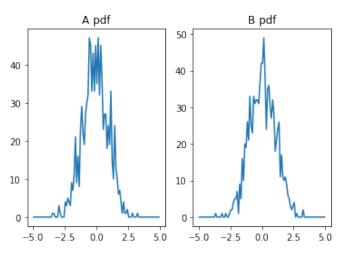


Figure 7: Sum of 2 probability mass function

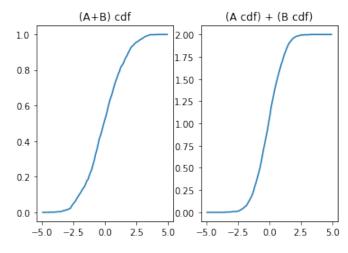


Figure 10: Sum of 2 cummulative density function

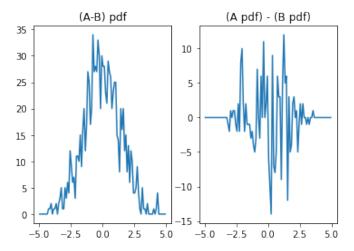


Figure 8: Subtract of 2 probability mass function

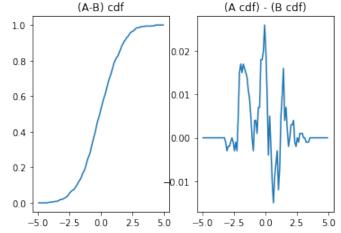


Figure 11: Subtract of 2 cummulative density function

So, basically normal distributions are approximation of binomial distributions.

PDF of 2 individual binomial random variables are plotted.

PDF of sum of 2 individual random variables and sum of their individual pdfs are plotted and compared.

PDF of difference of 2 individual random variables and difference of their individual pdfs are plotted and compared.

CDF of 2 individual binomial random variables are plotted.

CDF of sum of 2 individual random variables and sum of their individual Cdfs are plotted and compared.

CDF of difference of 2 individual random variables and difference of their individual cdfs are plotted and compared.

If x is a random variable with distribution Bin(n, p), then for sufficiently large n, the distribution of the variable.