

10th Material Subject: Special Distribution of Discrete Random Variable

Undergraduate of Telecommunication Engineering

MUH1F3 - PROBABILITY AND STATISTICS

Telkom University
Center of eLearning & Open Education Telkom University
Jl. Telekomunikasi No.1, Bandung - Indonesia
<http://www.telkomuniversity.ac.id>

Lecturer: Nor Kumalasari Caecar Pratiwi, S.T., M.T. (caecarnkcp@telkomuniversity.ac.id)



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TABLE OF CONTENTS:

1. **Bernoulli**
2. **Binomial**
3. **Hyper-geometric**
4. **Poisson**

LEARNING OBJECTIVES:

After careful study of this chapter, student should be able to do the following:

1. **Understand the assumptions for some common discrete probability distributions**
 2. **Calculate discrete probability distribution to calculate probabilities in specific applications**
 3. **Calculate probabilities and determine means and variances for some common discrete probability distributions**
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A trial with only two possible outcomes is used so frequently as a building block of a random experiment that it is called a **Bernoulli** trial.

- A Bernoulli distribution outcomes can only take two values, either **1 (success/yes)** or **0 (failure/no)**.
- It is usual to denote the probabilities by **p** and **q**, **p** as "**success**" and **q** as "**failure**", **p** and **q** must be non-negative, and their sum must be equal to one, **q = 1 - p**.
- Suppose **X** is Bernoulli distribution of probability success **p**, we can write:

$$X \Rightarrow \text{BER}(p) \quad (1)$$

- The **Probability Mass Function** of Bernoulli Distribution:

$$f_X(x) = \begin{cases} p & , x = 1 \\ q & , x = 0 \\ 0 & , \text{otherwise} \end{cases} \quad (2)$$

- The Me

$$\mathbf{E}(\mathbf{X}) = \sum \mathbf{x} \cdot \mathbf{f}(\mathbf{x}) = (\mathbf{1} \cdot \mathbf{p}) + (\mathbf{0} \cdot \mathbf{1}) = \mathbf{p} \quad (3)$$

- The **Variance** of Bernoulli Distribution:

$$\mathbf{E}(\mathbf{X}^2) = \sum \mathbf{x}^2 \cdot \mathbf{f}_\mathbf{X}(\mathbf{x}) = (1^2 \cdot \mathbf{p}) + (0^2 \cdot 1) = \mathbf{p}$$

$$\text{Var}(\mathbf{X}) = \sigma_{\mathbf{X}}^2 = \mathbf{E}(\mathbf{X}^2) - (\mathbf{E}(\mathbf{X}))^2 = \mathbf{p} - \mathbf{p}^2 = \mathbf{p}(1 - \mathbf{p}) = \mathbf{p}\mathbf{q} \quad (4)$$

- The **Moment Generation Function** of Bernoulli Distribution:

$$\mathbf{E}(\mathbf{e}^{\mathbf{t}\mathbf{x}}) = \sum \mathbf{e}^{\mathbf{t}\mathbf{x}} \cdot \mathbf{f}_{\mathbf{x}}(\mathbf{x}) = (\mathbf{e}^{\mathbf{t} \cdot 1} \cdot \mathbf{p}) + (\mathbf{e}^{\mathbf{t} \cdot 0} \cdot \mathbf{q}) = \mathbf{e}^{\mathbf{t}} \mathbf{p} + \mathbf{q} \quad (5)$$

Example: An experiment was made in choosing a students randomly out of 10 students (consist of 4 female and 6 male). If X is a random variable that is states the selection of female students, specify:

- Probability Mass Function (PMF) of random variable X
- Mean / Expected Value random variable X
- Variance random variable X
- Moment Generation Function random variable X

Answer: Probability for success p states if the student chosen is a female, so that $p = \frac{4}{10}$, create the $q = 1 - p = \frac{6}{10}$.

- Finally, the PMF can be written:

$$f_X(x) = \begin{cases} \frac{4}{10} & , x = 1 \\ \frac{6}{10} & , x = 0 \\ 0 & , \text{otherwise} \end{cases}$$

(6)

b. Mean / Expected Value of random variable **X**

$$E(X) = p = \frac{4}{10}$$

c. Variance of random variable **X**

$$\text{Var}(X) = pq = \frac{4}{10} \cdot \frac{6}{10} = \frac{24}{100}$$

d. Moment Generation Function of random variable **X**

$$E(e^{tx}) = e^t p + q = \frac{4}{10} e^t + \frac{6}{10}$$

Binomial is a random experiment that consists of **n** Bernoulli trials.

- A Binomial distribution outcomes can only take two values, either **1 (success/yes)** or **0 (failure/no)**.
- It is usual to denote the probabilities by **p** and **q**, **p** as "**success**" and **q** as "**failure**", **p** and **q** must be non-negative, and their sum must be equal to one, **q = 1 - p**.
- Suppose **X** is Binomial distribution with **n** trial and probability success **p**, we can write:

$$X \Rightarrow \text{BIN}(n, p) \quad (7)$$

- The **Probability Mass Function** of Binomial Distribution:

$$f_X(x) = \begin{cases} C_x^n \cdot p^x \cdot q^{n-x} & , x = 0, 1, 2, \dots, n \\ 0 & , \text{otherwise} \end{cases} \quad (8)$$

- The **Mean** of Binomial Distribution:

$$E(X) = np \quad (9)$$

- The **Variance** of Binomial Distribution:

$$\text{Var}(X) = npq \quad (10)$$

- The **Moment Generation Function** of Binomial Distribution:

$$E(e^{tx}) = (e^t p + q)^n \quad (11)$$

b. Mean / Expected Value of random variable **X**

$$E(X) = n \cdot p = 8 \cdot \frac{4}{10} = \frac{32}{10}$$

c. Variance of random variable **X**

$$\text{Var}(X) = n \cdot p \cdot q = 8 \cdot \frac{4}{10} \cdot \frac{6}{10} = \frac{192}{100}$$

d. Moment Generation Function of random variable **X**

$$E(e^{tx}) = (e^t p + q)^n = \left(\frac{4}{10} e^t + \frac{6}{10} \right)^8$$

Hypergeometric experiments have the same characteristics with binomial. The difference is hypergeometric states the number of successful events in samples taken as WOR (**Without Replacement**) samples.

- A set of N objects contains K objects classified as successes $N - K$ objects classified as failures. A sample of size n objects is selected randomly (without replacement) from the N objects where $K \leq N$ and $n \leq N$. The random variable X that equals the number of successes in the sample is a hypergeometric random.

$$X \Rightarrow \text{HYP}(N, n, K) \quad (13)$$

- The **Probability Mass Function** of Hypergeometric Distribution:

$$f_X(x) = \begin{cases} \frac{C_x^K \cdot C_{n-x}^{N-K}}{C_n^N} & , x = 0, 1, 2, \dots, n \\ 0 & , \text{otherwise} \end{cases} \quad (14)$$

(15)

- The **Mean** of Hypergeometric Distribution:

$$E(X) = \frac{n \cdot K}{N} = n \cdot p \quad (16)$$

- The **Variance** of Hypergeometric Distribution:

$$\text{Var}(X) = n \cdot p \cdot q \cdot \left(\frac{N - n}{N - 1} \right) \quad (17)$$

(18)

a. Probability Mass Function (PMF) random variable **X**

$$f_X(x) = \begin{cases} \frac{C_x^{10} \cdot C_{5-x}^6}{C_5^{16}} & , x = 0, 1, 2, 3, 4, 5 \\ 0 & , \text{otherwise} \end{cases}$$

b. Mean / Expected Value random variable **X**

$$E(X) = \frac{n \cdot K}{N} = n \cdot p = 5 \cdot \frac{5}{8} = \frac{25}{8}$$

c. Variance random variable **X**

$$\text{Var}(X) = n \cdot p \cdot q \cdot \left(\frac{N - n}{N - 1} \right) = 5 \cdot \frac{5}{8} \cdot \frac{3}{8} \cdot \frac{11}{15} = \frac{55}{64}$$

d. $f_X(0)$, $f_X(1)$, $f_X(2)$, $f_X(3)$, $f_X(4)$ and $f_X(5)$

$$f_X(0) = \frac{C_0^{10} \cdot C_{5-0}^6}{C_5^{16}} = \frac{1 \cdot 6}{4368} = \frac{6}{4368}$$

$$f_X(1) = \frac{C_1^{10} \cdot C_{5-1}^6}{C_5^{16}} = \frac{10 \cdot 15}{4368} = \frac{150}{4368}$$

$$f_X(2) = \frac{C_2^{10} \cdot C_{5-2}^6}{C_5^{16}} = \frac{45 \cdot 20}{4368} = \frac{900}{4368}$$

$$f_X(3) = \frac{C_3^{10} \cdot C_{5-3}^6}{C_5^{16}} = \frac{120 \cdot 15}{4368} = \frac{1800}{4368}$$

$$f_X(4) = \frac{C_4^{10} \cdot C_{5-4}^6}{C_5^{16}} = \frac{210 \cdot 6}{4368} = \frac{1260}{4368}$$

$$f_X(5) = \frac{C_5^{10} \cdot C_{5-5}^6}{C_5^{16}} = \frac{252 \cdot 1}{4368} = \frac{252}{4368}$$

Poisson first introduced by Sim eon-Denis Poisson(1781 - 1840). The Poisson distribution states the number of discrete events (sometimes also called "**arrivals**") that occur during certain time intervals.

- The random variable **X** that equals the number of events in a Poisson process is a poisson random variable with parameter $\lambda > 0$:

$$\mathbf{X} \Rightarrow \mathbf{POI}(\lambda) \quad (19)$$

- The **Probability Mass Function** of Poisson Distribution:

$$\mathbf{f_x(x)} = \begin{cases} \frac{e^{-\lambda} \cdot \lambda^x}{x!} & , x = 0, 1, 2, \dots \\ 0 & , \text{otherwise} \end{cases} \quad (20)$$

- The **Mean** of Poisson Distribution:

$$E(X) = \lambda \quad (21)$$

- The **Variance** of Poisson Distribution:

$$\text{Var}(X) = \lambda \quad (22)$$

- The **Moment Generation Function** of Poisson Distribution:

$$E(e^{tx}) = e^{\lambda(e^t - 1)} \quad (23)$$

Example: Phone calls that enter the complaint center meet the Poisson process with an entrance average of 2.5 call / minute. Calculate:

- Probability Mass Function (PMF) of random variable X
- Mean / Expected Value of random variable X
- Variance of random variable X
- Moment Generation Function of random variable X
- The probability of no more than 3 incoming calls in one minutes
- The probability of less than 5 incoming calls in one minutes

a. Probability Mass Function (PMF) random variable **X**

$$\mathbf{f}_{\mathbf{x}}(\mathbf{x}) = \begin{cases} \frac{e^{-2.5 \cdot 2.5^x}}{x!} & , x = 0, 1, 2, 3, \dots \\ 0 & , \text{otherwise} \end{cases}$$

b. Mean / Expected Value random variable **X**

$$E(X) = \lambda = 2.5$$

c. Variance random variable **X**

$$\text{Var}(\mathbf{X}) = \lambda = 2.5$$

d. Moment Generation Function of random variable **X**

$$E(e^{tx}) = e^{2.5(e^t - 1)} \quad (24)$$

e. The probability of no more than 3 incoming calls in one minutes

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

$$P(X \leq 3) = \frac{e^{-2.5} \cdot 2.5^0}{0!} + \frac{e^{-2.5} \cdot 2.5^1}{1!} + \frac{e^{-2.5} \cdot 2.5^2}{2!} + \frac{e^{-2.5} \cdot 2.5^3}{3!} = 0.7576$$



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Thank You