

Statistical Methods 105

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Chapter 1

Discrete Random Variable

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- 2 Some Discrete Probability Distributions
 - Discrete Uniform Random Variable
 - Binomial Distribution
 - Hypergeometric Distribution
 - Poisson Distribution

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1) Discrete Probability Distributions

Definition (Probability function)

The set of ordered pairs $(x, f(x))$ is a probability function, probability mass function, or probability distribution of the discrete random variable X if, for each possible outcome x ,

- ① $f(x) \geq 0$,
- ② $\sum_{x \in X} f(x) = 1$,
- ③ $P(X = x) = f(x)$.

Definition (cumulative distribution function)

The cumulative distribution function $F(x)$ of a discrete random variable X with probability distribution $f(x)$ is

$$F(x) = P(X \leq x) = \sum_{t \leq x} f(t), \text{ for } -\infty < x < +\infty.$$

Definition (Mean of a Random Variable)

Let X be a random variable with probability distribution $f(x)$. The mean, or expected value, of X is

$$\mu = E(x) = \sum_x x f(x).$$

Theorem

Let X be a random variable with probability distribution $f(x)$. The expected value of the random variable $g(X)$ is

$$\mu_{g(X)} = E[g(X)] = \sum_x g(x) f(x).$$

Example

Suppose that the number of cars X that pass through a car wash between 4:00 P.M. and 5:00 P.M. on any sunny Friday has the following probability distribution:

x	4	5	6	7	8	9
$f(x)$	$\frac{1}{12}$	$\frac{1}{12}$	$\frac{1}{4}$	$\frac{1}{4}$	$\frac{1}{6}$	$\frac{1}{6}$

1. Find $E(X)$.
2. Let $g(X) = 2X - 1$ represent the amount of money, in dollars, paid to the attendant by the manager. Find the attendant's expected earnings for this particular time period.

Solution

Simple calculations yield:

x	4	5	6	7	8	9
$f(x)$	$\frac{1}{12}$	$\frac{1}{12}$	$\frac{1}{4}$	$\frac{1}{4}$	$\frac{1}{6}$	$\frac{1}{6}$
$xf(x)$	$\frac{1}{3}$	$\frac{5}{12}$	$\frac{3}{2}$	$\frac{7}{4}$	$\frac{4}{3}$	$\frac{3}{2}$
$g(x)$	7	9	11	13	15	17
$g(x)f(x)$	$\frac{7}{12}$	$\frac{9}{12}$	$\frac{11}{4}$	$\frac{13}{4}$	$\frac{15}{6}$	$\frac{17}{6}$

1. $E(X) = \frac{1}{3} + \frac{5}{12} + \frac{3}{2} + \frac{7}{4} + \frac{4}{3} + \frac{3}{2} = 6.83$
2. The attendant's expected earnings for this particular time period is equal to:

$$E[g(X)] = \frac{9}{12} + \frac{11}{12} + \frac{13}{4} + \frac{15}{4} + \frac{17}{6} + \frac{19}{6} = 14.67.$$

Properties of the mean:

Theorem

Let X a random variable. If a and b are constants, then
 $E(aX + b) = aE(X) + b$.

Theorem

The expected value of the sum or difference of two or more functions of a random variable X is the sum or difference of the expected values of the functions. That is,

$$E[g(X) \pm h(X)] = E[g(X)] \pm E[h(X)].$$

Example

Let X be a random variable with probability distribution as follows:

x	0	1	2	3
$f(x)$	$\frac{1}{3}$	$\frac{1}{2}$	0	$\frac{1}{6}$

Find the expected value of $Y = (X - 1)^2$.

Solution

Simple calculations yield:

x	0	1	2	3
$f(x)$	$\frac{1}{3}$	$\frac{1}{2}$	0	$\frac{1}{6}$
$g(x)$	1	0	1	4
$f(x)g(x)$	$\frac{1}{3}$	0	0	$\frac{2}{3}$

Therefore, the expected value of Y is equal to:

$$E(Y) = E[g(X)] = 1.$$

Another solution: by using the properties of the mean theorems,

$$E(Y) = E((X - 1)^2) = E(X^2) - 2E(X) + 1!$$

Theorems (Variance of Random Variable)

Let X be a random variable with probability distribution $f(x)$ and mean μ . The variance of X is

$$\text{Var}(X) = \sigma^2 = E[(X - \mu)^2] = \sum_x (x - \mu)^2 f(x).$$

or it can be written as:

$$\sigma^2 = E(X^2) - E(X)^2$$

$$\text{Var}(g(X)) = \sigma_{g(X)}^2 = E[(g(x) - \mu_{g(X)})^2] = \sum_x (g(x) - \mu_{g(X)})^2 f(x).$$

The positive square root of the variance, σ , is called the standard deviation of X .

Properties of the Variance:

Theorem

Let X a random variable. If a and b are constants, then $\text{Var}(aX + b) = a^2 \text{Var}(X)$.

Corollary

Setting $a = 1$, then $\text{Var}(X + b) = \text{Var}(X)$.

Corollary

Setting $b = 0$, then $\text{Var}(aX) = a^2 \text{Var}(X)$.

Example

Calculate the variance of $g(X) = 2X + 3$, where X is a random variable with probability distribution:

x	0	1	2	3
$f(x)$	$\frac{1}{4}$	$\frac{1}{8}$	$\frac{1}{2}$	$\frac{1}{8}$

Solution

Simple calculations yield

x	0	1	2	3
$f(x)$	$\frac{1}{4}$	$\frac{1}{8}$	$\frac{1}{2}$	$\frac{1}{8}$
$g(x)$	3	5	7	9
$g(x)f(x)$	$\frac{3}{4}$	$\frac{5}{8}$	$\frac{7}{2}$	$\frac{9}{8}$

Therefore, The expected value of $g(X)$ is equal to

$$E[g(X)] = \frac{3}{4} + \frac{5}{8} + \frac{7}{2} + \frac{9}{8} = 6$$

So, the variance of $g(X) = 2X + 3$ is equal to

$$\sigma^2 = (3-6)^2 * \frac{1}{4} + (5-6)^2 * \frac{1}{8} + (7-6)^2 * \frac{1}{2} + (9-6)^2 * \frac{1}{8} = 4,$$

and the standard deviation of $g(X)$ is equal to: $\sigma = \sqrt{4} = 2$.

Another solution: By using the properties of the variance

$$\text{Var}(2X + 3) = 2^2 \text{Var}(X) = 4 \text{Var}(X) = 4[E(X^2) - E(X)^2]$$

x	0	1	2	3
$f(x)$	$\frac{1}{4}$	$\frac{1}{8}$	$\frac{1}{2}$	$\frac{1}{8}$
x^2	0	1	4	9

Therefore, $E[X^2] = \frac{1}{8} + 2 + \frac{9}{8} = \frac{26}{8}$, and $E[X] = \frac{1}{8} + 1 + \frac{3}{8} = \frac{12}{8}$.

Then, the variance of $g(X) = 2X + 3$ is equal to

$$\sigma^2 = 4 \left[\frac{26}{8} - \left(\frac{12}{8} \right)^2 \right] = 4.$$

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2.1) Discrete Uniform Random Variable

Definition (Discrete Uniform Random Variable)

A random variable X is called discrete uniform if has a finite number of specified outcomes, say x_1, x_2, \dots, x_k and each outcome is equally likely. Then, the discrete uniform mass function is given by:

$$f(x) = P(X = x) = \begin{cases} \frac{1}{k}, & x = x_1, x_2, \dots, x_k \\ 0, & \text{otherwise.} \end{cases}$$

Note: k is called the parameter of the distribution.

Theorem

The expected value (mean) and variance of the discrete uniform distribution are:

$$\mu = E(X) = \sum_{i=1}^k \frac{x_i}{k}, \text{ and } \sigma^2 = \frac{1}{k} \sum_{i=1}^k [x_i - E(X)]^2.$$

Example

Suppose that you select a ball from a box contains 6 balls labeled $1, 2, \dots, 6$. Let X = the number that is observed when selecting a ball. Find $E(X)$ and $Var(X)$.

Solution

The probability distribution of X is:

$$P(X = x) = \begin{cases} \frac{1}{6}, & x = x_1, x_2, \dots, x_6 \\ 0, & \text{otherwise.} \end{cases}$$

The expected value:

$$\mu = E(X) = \sum_{i=1}^k \frac{x_i}{k} = \frac{1 + 2 + 3 + 4 + 5 + 6}{6} = 3.5.$$

The variance:

$$\sigma^2 = \frac{1}{k} \sum_{i=1}^k [x_i - E(X)]^2 = \frac{(1 - 3.5)^2 + \dots + (6 - 3.5)^2}{6} = 2.92.$$

2.2) Binomial Distribution

Definition (Bernoulli Process)

The process is said to be a Bernoulli process if:

- The outcomes of process is either success or failure.
- The probability of success is $P(X = 1) = p$ and the probability of failure is $P(X = 0) = 1 - p = q$.

Strictly speaking, trials of random experiment are called Bernoulli trials if satisfy the following conditions:

- ① The experiment consists of finite number of repeated trials.
- ② Each trial has exactly two outcomes: success or failure.
- ③ The repeated trials are independent.
- ④ The probability of success remains the same in each trial.

Definition (Binomial Distribution)

The binomial distribution is defined based on the Bernoulli process. It is made up of n independent Bernoulli processes. Suppose that X_1, X_2, \dots, X_n are independent Bernoulli random variables, then $Y = \sum X_i$ will conform binomial distribution. The probability mass function of the binomial random variable X is given by:

$$f(x) = P(X = x) = \binom{n}{x} p^x q^{n-x}, \quad x = 0, 1, 2, \dots, n.$$

We denote the binomial distribution with the parameters n and p , by $Bin(n, p)$ and $\binom{n}{x} = \frac{n!}{x!(n-x)!}$.

The Cumulative Distribution Function (CDF)

The cumulative distribution function (CDF) of binomial distribution $\text{Bin}(n, p)$ is:

$$F_X(x) = P(X \leq x) = \sum_{i=0}^x \binom{n}{i} p^i q^{n-i}.$$

Theorem

The mean and variance of the binomial distribution $\text{Bin}(n, p)$ are

$$\mu = np \text{ and } \sigma^2 = npq.$$

Example

If the mean and the variance of a binomial distribution are 10 and 5 respectively, then:

- 1 Determine the probability mass function.
- 2 Calculate the probability $P(X = 0)$.
- 3 Calculate the probability $P(X \geq 2)$.

Solution

- 1 By solving $E(X) = np = 10$ and $Var(X) = np(1 - p) = 5$, we get $p = 0.5$ and $n = 20$. The probability mass function is:

$$P(X = x) = \binom{20}{x} (0.5)^x (0.5)^{20-x}, \quad x = 0, 1, \dots, 20$$

- 2 $P(X = 0) = \binom{20}{0} (0.5)^0 (0.5)^{20} = 0.5^{20}$.
- 3 $P(X \geq 2) = 1 - P(X < 2) = 1 - [P(X = 0) + P(X = 1)] = 1 - 0.00002 = 0.99998$.

Example

Suppose that the probability that a person dies when he or she contracts a certain disease is 0.4. A sample of 10 persons who contracted this disease is randomly chosen. Find the following:

- 1 The probability that exactly 4 persons will die.
- 2 The probability that less than 3 persons will die.
- 3 The probability that more than 8 persons will die.
- 4 The expected number of persons who will die.
- 5 The variance of the number of persons who will die.

Solution

The probability mass function is:

$$P(X = x) = \binom{10}{x} (0.4)^x (0.6)^{10-x}, \quad x = 0, 1, \dots, 10$$

$$1. \quad P(X = 4) = \binom{10}{4} (0.4)^4 (0.6)^{10-4} = 0.251$$

2.

$$\begin{aligned} P(X < 3) &= P(X = 0) + P(X = 1) + P(X = 2) \\ &= \sum_{x=0}^2 \binom{10}{x} (0.4)^x (0.6)^{10-x} = 0.167. \end{aligned}$$

3.

$$\begin{aligned} P(X > 8) &= P(X = 9) + P(X = 10) \\ &= \sum_{x=9}^{10} \binom{10}{x} (0.4)^x (0.6)^{10-x} = 0.0017. \end{aligned}$$

4. $E(X) = np = (10)(0.4) = 4$

5. $Var(X) = np(1 - p) = (10)(0.4)(0.6) = 2.4$

2.3) Hypergeometric Distribution

Definition

The probability distribution of the hypergeometric random variable X describes the probability of K successes (random draws for which the object drawn has a specified feature) in n draws, without replacement, from a finite population of size N that contains exactly K objects with that feature, where each draw is either a success or a failure.

There are two methods of selection:

1. Selection with replacement: If we select the elements of the sample at random and with replacement, then $X \sim \text{Bin}(n, p)$; where $p = \frac{K}{N}$
2. Selection without replacement: When the selection is made without replacement, the random variable X has a hypergeometric distribution with parameters N, n , and K . and we write $X \sim h(x; N, n, K)$.

The probability mass function for hypergeometric random variable X is:

$$f(x) = P(X = x) = \begin{cases} \frac{\binom{K}{x} \binom{N-K}{n-x}}{\binom{N}{n}} & x = 0, 1, 2, \dots, \min(K, n) \\ 0, & \text{otherwise.} \end{cases}$$

Theorem

The mean and variance of the hypergeometric distribution $h(x; N, n, K)$ are

$$\mu = n \frac{K}{N} \text{ and } \sigma^2 = n \frac{K}{N} \left(1 - \frac{K}{N}\right) \frac{N-n}{N-1}.$$

Example

Suppose there are 50 officers, 10 female officers and 40 male officers. Suppose 20 of them will be selected for promotion. Let X represent the number of female promotions. Find:

- 1 The probability that exactly one female is found in the sample.
- 2 The expected value (mean) and the variance of the number of females in the sample.

Solution

- Note that the binomial distribution doesn't apply here, as the officers are without replacement once they are drawn. In other words, the trials are not independent events.
- X has a hypergeometric distribution with $N = 50$, $n = 20$, and $K = 10$; i.e. $X \sim h(x; N, n, K) = h(x; 50, 20, 10)$.

$$P(X = x) = \begin{cases} \frac{\binom{10}{x} \binom{50-10}{20-x}}{\binom{50}{20}} & x = 0, 1, 2, \dots, 10 \\ 0, & \text{otherwise.} \end{cases}$$

- ① The probability that exactly one female is found in the sample is:

$$f(1) = P(X = 1) = \frac{\binom{10}{1} \binom{40}{19}}{\binom{50}{20}} = 0.0279$$

- ② The expected value (mean) is $E(X) = n \frac{K}{N} = 20 \times \frac{10}{50} = 4$.

- ③ The variance is
- $$\sigma^2 = n \frac{K}{N} \left(1 - \frac{K}{N}\right) \frac{N-n}{N-1} = 20 \times \frac{10}{50} \times \left(1 - \frac{10}{50}\right) \times \frac{50-20}{50-1} = 1.9592$$

(Binomial Approximation Theorem)

If n is small compared to K , then a binomial distribution $Bin(n, p = \frac{K}{N})$ can be used to approximate the hypergeometric distribution $h(x; N, n, K)$.

Example

A manufacturer of automobile tires reports that among a shipment of 5000 sent to a local distributor, 1000 are slightly blemished. If one purchases 10 of these tires at random from the distributor, what is the probability that exactly 3 are blemished?

Solution

Since $K = 1000$ is large relative to the sample size $n = 10$, we shall approximate the desired probability by using the binomial distribution. The probability of obtaining a blemished tire is 0.2. Therefore, the probability of obtaining exactly 3 blemished tires is

$$h(3; 5000, 10, 1000) \approx Bin(10, p = \frac{1000}{5000}) = \binom{10}{3} (0.2)^3 (0.8)^7 = 0.2013.$$

2.4)Poisson Distribution

Definition

Let X the number of outcomes occurring during a given time interval. X is called a Poisson random variable, with parameter λ , when its probability mass function is given by

$$p(x, \lambda) = P(X = x) = e^{-\lambda} \frac{\lambda^x}{x!}, \quad x = 0, 1, 2, \dots$$

where e is an irrational number approximately equal to 2.71828 and λ is the average number of occurrences per interval unit.

Theorem

If a random variable X has a Poisson distribution. Then both the mean and the variance of X are λ .

$$\mu = \lambda \text{ and } \sigma^2 = \lambda$$

Example

The mean number of accidents per month at a certain intersection is 3.

- ① What is the probability that in any given month 4 accidents will occur at this intersection?
- ② What is the probability that more than 4 accidents will occur in any given month at the intersection?
- ③ What is the probability that 4 accidents will occur in 5 months?

Solution

- ① $f(4) = P(X = 4) = e^{-3} \frac{3^4}{4!} = 0.168.$
- ② $P(X > 4) = 1 - P(X \leq 4) = 1 - [P(X = 0) + \cdots + P(X = 4)] = 1 - [\sum_{x=0}^4 e^{-3} \frac{3^x}{x!}] = 0.1847.$
- ③ Since the average number of accidents at a certain intersection per month is 3, thus the average number of accidents in 5 months is 15. Let X represent the number of accidents in 5 months, $f(x) = P(X = x) = e^{-15} \frac{15^x}{x!}$, $x = 0, 1, 2, \dots$.
Then, $f(4) = P(X = 4) = e^{-15} \frac{15^4}{4!} = 0.00065.$

Theorem (Approximation)

Let X be a binomial random variable with probability distribution $B(n, p)$. When n is large ($n \rightarrow +\infty$), and p small ($p \rightarrow 0$), then the poisson distribution can be used to approximate the binomial distribution $B(n, p)$ by taking $\lambda = np$.

Example

In a certain industrial facility, accidents occur infrequently. It is known that the probability of an accident on any given day is 0.005 and accidents are independent of each other.

- 1 What is the probability that in any given period of 400 days there will be an accident on one day?
- 2 What is the probability that there are at most three days with an accident?

Solution

Let X be a binomial random variable with $n = 400$ and $p = 0.005$. Thus, $np = 2$. Use the Poisson approximation,

1

$$P(X = 1) = e^{-2} 2^1 = 0.271.$$

2

$$P(X \leq 3) = \sum_{x=0}^3 e^{-2} \frac{2^x}{x!} = 0.857.$$