

Statistical Methods 105

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Chapter 1

Discrete random variable

- 1 Discrete probability distributions
- 2 Some Discrete probability Distributions
 - Discrete Uniform Random Variable
 - Binomial Distribution
- 3 Hypergeometric Distribution
- 4 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

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

Example

A lot containing 7 components is sampled by a quality inspector; the lot contains 4 good components and 3 defective components. A sample of 3 is taken by the inspector. Find the expected value of the number of good components in this sample.

Solution

Assume X represents the number of good components in the sample. The probability distribution of X is

$$f(x) = \frac{\binom{4}{x} \binom{3}{3-x}}{\binom{7}{3}}, \quad x = 1, 2, 3. \quad (1)$$

Using the formula (1), we obtain

$$f(0) = \frac{1}{35}, f(1) = \frac{12}{35}, f(2) = \frac{18}{35} \text{ and } f(3) = \frac{4}{35}.$$

Therefore,

$$\mu = E(x) = 0 * f(0) + 1 * f(1) + 2 * f(2) + 3 * f(3) = \frac{12}{7} = 1.7.$$

Thus, if a sample of size 3 is selected at random over and over again from a lot of 4 good components and 3 defective components, it will contain, on average, 1.7 good components.

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
$f(x)$	$\frac{1}{12}$	$\frac{1}{12}$	$\frac{1}{4}$	$\frac{1}{4}$	$\frac{1}{6}$	$\frac{1}{6}$

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
$f(x)$	$\frac{1}{12}$	$\frac{1}{12}$	$\frac{1}{4}$	$\frac{1}{4}$	$\frac{1}{6}$	$\frac{1}{6}$
$g(x)$	9	11	13	15	17	19
$f(x)g(x)$	$\frac{9}{12}$	$\frac{11}{12}$	$\frac{13}{4}$	$\frac{15}{4}$	$\frac{17}{6}$	$\frac{19}{6}$

Therefore, the attendant's expected earnings for this particular time period is equal to:

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

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:

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

Theorems (Variance of Random Variable)

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

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

The positive square root of the variance, σ , is called the standard deviation of 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
$f(x)g(x)$	$\frac{3}{4}$	$\frac{5}{8}$	$\frac{7}{2}$	$\frac{9}{8}$

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: $\sigma = \sqrt{4} = 2$.

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2.1) Discrete Uniform Random Variable

Definition (Discrete Uniform Random Variable)

A random variable X is called discrete uniform if it has a finite number of possible values, say x_1, x_2, \dots, x_n and

$$P(X = x_i) = \begin{cases} \frac{1}{n}, & \text{for all } 1 \leq i \leq n \\ 0, & \text{elsewhere.} \end{cases}$$

Note: n is called the parameter of the distribution.

Example

Experiment: tossing a balanced die.

- Sample space: $S = \{1, 2, 3, 4, 5, 6\}$.
- Each sample point of S occurs with the same probability $\frac{1}{6}$.
- Let $X =$ the number observed when tossing a balanced die.

Solution

The probability distribution of X is:

$$P(X = x) = \begin{cases} \frac{1}{6}, & \text{for all } 1 \leq x \leq 6 \\ 0, & \text{elsewhere.} \end{cases}$$

Definition (Bernoulli Process)

Strictly speaking, the Bernoulli process must possess the following properties:

- 1 The experiment consists of repeated trials.
- 2 Each trial results in an outcome that may be classified as a success or a failure.
- 3 The probability of success, denoted by p , remains constant from trial to trial.
- 4 The repeated trials are independent.

Definition (Binomial Distribution)

A Bernoulli trial can result in a success with probability p and a failure with probability $q = 1 - p$. Then the probability distribution of the binomial random variable X , the number of successes in n independent trials, is

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

Example

The probability that a certain kind of component will survive a shock test is $\frac{3}{4}$. Find the probability that exactly 2 of the next 4 components tested survive.

Solution

Let X the number of components that will survive a shock test.

Assuming that the tests are independent and $p = \frac{3}{4}$ for each of the 4 tests, then X is a binomial distribution Binomial($4, \frac{3}{4}$) or $B(4, \frac{3}{4})$.

Hence,

$$P(X = 2) = \binom{4}{2} \left(\frac{3}{4}\right)^2 \left(\frac{1}{4}\right)^2 \approx 0.21,$$

and

$$P(X = 0) = \binom{4}{0} \left(\frac{3}{4}\right)^0 \left(\frac{1}{4}\right)^4 = 0.0625,$$

and

$$P(X = 5) = 0.$$

Example

The probability that a patient recovers from a rare blood disease is 0.4. If 15 people are known to have contracted this disease, what is the probability that

- 1 at least 10 survive,
- 2 from 3 to 8 survive,
- 3 exactly 5 survive?

Solution

1)

$$\begin{aligned}P(X \geq 10) &= 1 - P(X < 10) \\&= 1 - \sum_{x=0}^9 \binom{15}{x} (0.4)^x (0.6)^{15-x} \\&= 1 - 0.9662 = 0.0338\end{aligned}$$

2)

$$P(3 \leq X \leq 8) = \sum_{x=3}^8 \binom{15}{x} (0.4)^x (0.6)^{15-x} = 0.8779.$$

3)

$$P(X = 5) = \binom{15}{5} (0.4)^5 (0.6)^{15-5} = 0.1859.$$

Theorem

The mean and variance of the binomial distribution $B(n, p)$ are

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

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3) Hypergeometric Distribution

Definition

The probability distribution of the hypergeometric random variable X , the number of successes in a random sample of size n selected from N items of which K are labeled success and $N - K$ labeled failure, is

$$h = (x, N, n, K) = P(X = x) = \frac{\binom{K}{x} \binom{N-K}{n-x}}{\binom{N}{n}}.$$

Theorem

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

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

Example

Lots of 40 components each are deemed unacceptable if they contain 3 or more defectives. The procedure for sampling a lot is to select 5 components at random and to reject the lot if a defective is found. What is the probability that exactly 1 defective is found in the sample if there are 3 defectives in the entire lot?

Solution

Using the hypergeometric distribution with $n = 5$, $N = 40$, $k = 3$, and $x = 1$, we find the probability of obtaining 1 defective to:

$$h(1, 40, 5, 3) = \frac{\binom{3}{1} \binom{37}{4}}{\binom{40}{5}} = 0.3011.$$

Theorem (Approximation)

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

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

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

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3) 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 distribution is given by

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

where λ is the average number of outcomes.

Example

During a laboratory experiment, the average number of radioactive particles passing through a counter in 1 millisecond is 4. What is the probability that 6 particles enter the counter in a given millisecond?

Solution

Using the Poisson distribution with $x = 6$ and $\lambda = 4$, we have

$$p(6, 4) = e^{-4} \frac{4^6}{6!} = 0.1041.$$

Theorem

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

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

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

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