

MAT 2377 3X (Spring 2011)

§3.5 Discrete Uniform Distribution

Definition : A random variable has a discrete uniform distribution if it has a finite range $R_X = \{x_1, \dots, x_n\}$ and that

$$f(x_i) = \frac{1}{n}, \quad \text{pour } i = 1, \dots, n.$$

Special Case : If X has a discrete uniform distribution on the consecutive integers $a, a + 1, a + 2, \dots, b$, then

$$\mu = E[X] = \frac{a + b}{2}$$

and

$$\sigma^2 = V(X) = \frac{(b - a + 1)^2 - 1}{12}.$$

Example 5 : Suppose that the number of employees that are absent for medical reasons in the month of January has a discrete uniform distribution on 0 to 5 employees. Let X the number of employees that are absent due to medical reasons in January. Compute the mean and the standard deviation of X .

§3.6 Binomial Distribution

Définition : A *Bernoulli trial* is a random experiment with two possible outcomes that we will call “success” and “failure”. The probability that a success will occur is p .

Definition : A *binomial experiment* consists of n independent Bernoulli trials, each having the same probability of success.

Definition : Let X the number of successes for a binomial experiment of n Bernoulli trials, each with a probability p of success. We say that X has a binomial distribution with parameters n and p .

1. **Notation :** $X \sim B(n, p)$ means that X has a binomial distribution with the parameters n and p .
2. The p.m.f. for a binomial random variable is

$$f(x) = P(X = x) = \binom{n}{x} p^x (1 - p)^{n-x},$$

for $x = 0, 1, \dots, n$

3. $\mu = E[X] = np$ and $\sigma^2 = V(X) = np(1 - p)$
4. The c.d.f. for $X \sim B(n, p)$ is found in Table II for certain values n and p .

Example 6 : Example 1 : During one stage in the manufacture of integrated circuit chips, a coating must be applied. If 70% of chips receive a thick enough coating, then compute the probabilities that, among 15 chips :

- (a) exactly 10 will have thick enough coatings ;
- (b) at most 2 will have thick enough coatings ;
- (c) at least 14 will have thick enough coatings ;

Example 7 : Consider the random experiment from Example 6.

- (a) What is the expected number of chips with a thick enough coating among 15 chips ?
- (b) Compute the standard deviation of the number of chips with a thick enough coating among 15 chips ?

§3.7.1 Geometric Distribution

Definition : Let X be the number of independent Bernoulli trials required to observe a success. We assume that each trial has the same probability of success. We say that X has a geometric distribution with parameter p .

1. The p.m.f. for the geometric distribution is

$$f(x) = P(X = x) = (1 - p)^{x-1} p, \quad \text{for } x = 1, 2, 3, \dots$$

2. Its mean is

$$\mu = E(X) = \frac{1}{p}$$

3. Its variance is

$$\sigma^2 = V(X) = \frac{1 - p}{p^2}$$

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$$\begin{aligned} F(x) &= P(X \leq x) \\ &= 1 - P(X > x) \\ &= 1 - P(\text{"the first } x \text{ trials are failures" }) \\ &= 1 - (1 - p)^x, \end{aligned}$$

for $x = 1, 2, 3, \dots$

Example 8 : Electronic components are shipped to a supplier in boxes of 10. The supplier will return a box to the manufacturer only if it contains more than one defective component. Suppose that 5% of the components are defective.

- (a) What is the probability that the 10th box received by the supplier will be the first that is returned to the manufacturer?
- (b) What is the expected number of boxes that the manufacturer will have received when it first returns a box to the supplier?
- (c) What is the probability that more than 10 boxes will be received by the supplier before the first box is returned?

§3.7.2 Negative Binomial

Definition : Let X be the number of independent Bernoulli trials required to observe r successes. Each trial has the same probability of success. We say that X has a negative binomial distribution with parameters p and r .

1. The p.m.f. of the negative binomial distribution is

$$f(x) = P(X = x) = \binom{x-1}{r-1} (1-p)^{x-r} p^r,$$

for $x = r, r+1, r+2, \dots$

2. Its mean is

$$\mu = E(X) = \frac{r}{p}$$

3. Its variance is

$$\sigma^2 = V(X) = \frac{r(1-p)}{p^2}$$

Example 9 : Electronic components are shipped to a supplier in boxes of 10. The supplier will return a box to the manufacturer only if it contains more than one defective component. Suppose that 2.5% of boxes contain more than one defective component. What is the probability that the 10th box received by the supplier will be the 3rd that is returned to the manufacturer?

§3.9 The Poisson Distribution

Definition : A random variable X is said to have a Poisson distribution with parameter $\lambda > 0$ if it has the following probability mass function :

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

Mean and Variance : The mean and variance of a Poisson random variable X are equal to the parameter λ , that is

$$\mu = E[X] = \lambda \quad \text{and} \quad \sigma^2 = V(X) = \lambda.$$

Poisson approximation to the binomial distribution :

If X is a binomial random variable with n large and p small then the binomial p.m.f. may be approximated by the Poisson p.m.f. with $\lambda = np$:

$$P(X = x) = \binom{n}{x} p^x (1-p)^{n-x} \approx e^{-\lambda} \frac{\lambda^x}{x!},$$

where $\lambda = np$.

Rule of thumb :

The approximation is good, if $n \geq 20$ and $p \leq 0.05$.

It is very good if $n \geq 100$ and $np \leq 10$.

Example 10 : In a given city, 3% of all drivers get at least one parking ticket per year. Approximate the probabilities that among 80 drivers (randomly chosen in this city) :

- (a) 3 will get at least one parking ticket in any given year ;
- (b) at least 3 will get at least one parking ticket in any given year ;
- (c) anywhere from 2 to 5, inclusive, will get at least one parking ticket in any given year.

Poisson Process

Note : We have already seen that the Poisson Distribution can be used to approximate a binomial distribution when n is large and p is small. Below we will encounter another application of the Poisson distribution.

Definition : Let the number of “changes” that occur in a continuous interval (of time or space) be counted. We have a Poisson process of rate α if :

1. The number of changes occurring in nonoverlapping intervals are independent.
2. The probability of exactly one change in a short interval of length h is approximately αh .
3. The probability of two or more changes in a sufficiently short interval is essentially 0.

Counting “changes” in an interval of fixed length : Let X be the number of changes in an interval of length t in a Poisson process of rate α . Then, X has a Poisson distribution with parameter λ , where $\lambda = \alpha t$.

Example 11 : Assume that the calls to a call center occur according a Poisson process. If on average there are 6 calls per hour, what is the probability that

- (a) there are no calls in one hour ?
- (b) there are at most 2 calls in 30 minutes ?
- (c) there are exactly 10 calls in 2 hours ?

Example 12 : Errors on a data tape occur at a rate of one error per 35,000 centimeters. Assume that errors occur as a Poisson process.

- (a) Give the distribution of X , where X is the number of errors in 52,500 centimeters ?
- (b) Compute the probability that there is at least on error in 52 500 centimeters ?