

Chapter 5: Some Discrete Probability Distributions:

5.2: Discrete Uniform Distribution:

If the discrete random variable X assumes the values x_1, x_2, \dots, x_k with equal probabilities, then X has the discrete uniform distribution given by:

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

Note:

- $f(x) = f(x; k) = P(X = x)$
- k is called the parameter of the distribution.

Example 5.2:

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

$$f(x) = P(X = x) = f(x; 6) = \begin{cases} \frac{1}{6} & ; x = 1, 2, \dots, 6 \\ 0 & ; elsewhere \end{cases}$$

Theorem 5.1:

If the discrete random variable X has a discrete uniform distribution with parameter k , then the mean and the variance of X are:

$$E(X) = \mu = \frac{\sum_{i=1}^k x_i}{k}$$

$$\text{Var}(X) = \sigma^2 = \frac{\sum_{i=1}^k (x_i - \mu)^2}{k}$$

Example 5.3:

Find $E(X)$ and $\text{Var}(X)$ in Example 5.2.

Solution:

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

$$\begin{aligned} \text{Var}(X) = \sigma^2 &= \frac{\sum_{i=1}^k (x_i - \mu)^2}{k} = \frac{\sum_{i=1}^k (x_i - 3.5)^2}{6} \\ &= \frac{(1-3.5)^2 + (2-3.5)^2 + \dots + (6-3.5)^2}{6} = \frac{35}{12} \end{aligned}$$

5.3 Binomial Distribution:

Bernoulli Trial:

- Bernoulli trial is an experiment with only two possible outcomes.
- The two possible outcomes are labeled: success (s) and failure (f)
- The probability of success is $P(s)=p$ and the probability of failure is $P(f)=q = 1-p$.
- Examples:
 1. Tossing a coin (success= H , failure= T , and $p=P(H)$)
 2. Inspecting an item (success= defective , failure= non-defective , and $p=P(\text{defective})$)

Bernoulli Process:

Bernoulli process is an experiment that must satisfy the following properties:

1. The experiment consists of n repeated Bernoulli trials.
2. The probability of success, $P(s)=p$, remains constant from trial to trial.
3. The repeated trials are independent; that is the outcome of one trial has no effect on the outcome of any other trial

Binomial Random Variable:

Consider the random variable :

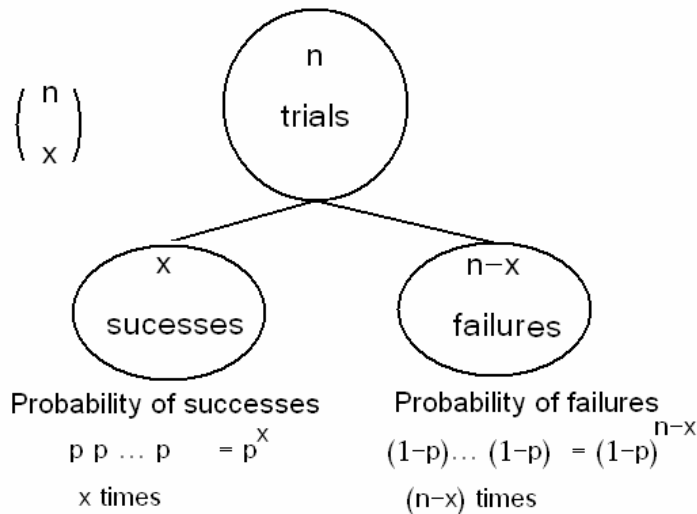
X = The number of successes in the n trials in a Bernoulli process

The random variable X has a binomial distribution with parameters n (number of trials) and p (probability of success), and we write:

$$X \sim \text{Binomial}(n,p) \text{ or } X \sim b(x;n,p)$$

The probability distribution of X is given by:

$$f(x) = P(X = x) = b(x;n,p) = \begin{cases} \binom{n}{x} p^x (1-p)^{n-x}; & x=0, 1, 2, \dots, n \\ 0; & \text{otherwise} \end{cases}$$



We can write the probability distribution of X as a table as follows.

x	$f(x)=P(X=x)=b(x;n,p)$
0	$\binom{n}{0} p^0 (1-p)^{n-0} = (1-p)^n$
1	$\binom{n}{1} p^1 (1-p)^{n-1}$
2	$\binom{n}{2} p^2 (1-p)^{n-2}$
\vdots	\vdots
$n-1$	$\binom{n}{n-1} p^{n-1} (1-p)^1$
n	$\binom{n}{n} p^n (1-p)^0 = p^n$
Total	1.00

Example:

Suppose that 25% of the products of a manufacturing process are defective. Three items are selected at random, inspected, and classified as defective (D) or non-defective (N). Find the probability distribution of the number of defective items.

Solution:

- Experiment: selecting 3 items at random, inspected, and classified as (D) or (N).
- The sample space is

$$S = \{DDD, DDN, DND, DNN, NDD, NDN, NND, NNN\}$$
- Let X = the number of defective items in the sample
- We need to find the probability distribution of X .

(1) First Solution:

Outcome	Probability	x
NNN	$\frac{3}{4} \times \frac{3}{4} \times \frac{3}{4} = \frac{27}{64}$	0
NND	$\frac{3}{4} \times \frac{3}{4} \times \frac{1}{4} = \frac{9}{64}$	1
NDN	$\frac{3}{4} \times \frac{1}{4} \times \frac{3}{4} = \frac{9}{64}$	1
NDD	$\frac{3}{4} \times \frac{1}{4} \times \frac{1}{4} = \frac{3}{64}$	2
DNN	$\frac{1}{4} \times \frac{3}{4} \times \frac{3}{4} = \frac{9}{64}$	1
DND	$\frac{1}{4} \times \frac{3}{4} \times \frac{1}{4} = \frac{3}{64}$	2
DDN	$\frac{1}{4} \times \frac{1}{4} \times \frac{3}{4} = \frac{3}{64}$	2
DDD	$\frac{1}{4} \times \frac{1}{4} \times \frac{1}{4} = \frac{1}{64}$	3

The probability distribution
of X is

.x	.f(x)=P(X=x)
0	$\frac{27}{64}$
1	$\frac{9}{64} + \frac{9}{64} + \frac{9}{64} = \frac{27}{64}$
2	$\frac{3}{64} + \frac{3}{64} + \frac{3}{64} = \frac{9}{64}$
3	$\frac{1}{64}$

(2) Second Solution:

Bernoulli trial is the process of inspecting the item. The results are success=D or failure=N, with probability of success $P(s)=25/100=1/4=0.25$.

The experiments is a Bernoulli process with:

- number of trials: $n=3$
- Probability of success: $p=1/4=0.25$
- $X \sim \text{Binomial}(n,p)=\text{Binomial}(3,1/4)$

- The probability distribution of X is given by:

$$f(x) = P(X = x) = b(x; 3, \frac{1}{4}) = \begin{cases} \binom{3}{x} \left(\frac{1}{4}\right)^x \left(\frac{3}{4}\right)^{3-x}; & x = 0, 1, 2, 3 \\ 0; & \text{otherwise} \end{cases}$$

$$f(0) = P(X = 0) = b(0; 3, \frac{1}{4}) = \binom{3}{0} \left(\frac{1}{4}\right)^0 \left(\frac{3}{4}\right)^3 = \frac{27}{64}$$

$$f(1) = P(X = 1) = b(1; 3, \frac{1}{4}) = \binom{3}{1} \left(\frac{1}{4}\right)^1 \left(\frac{3}{4}\right)^2 = \frac{27}{64}$$

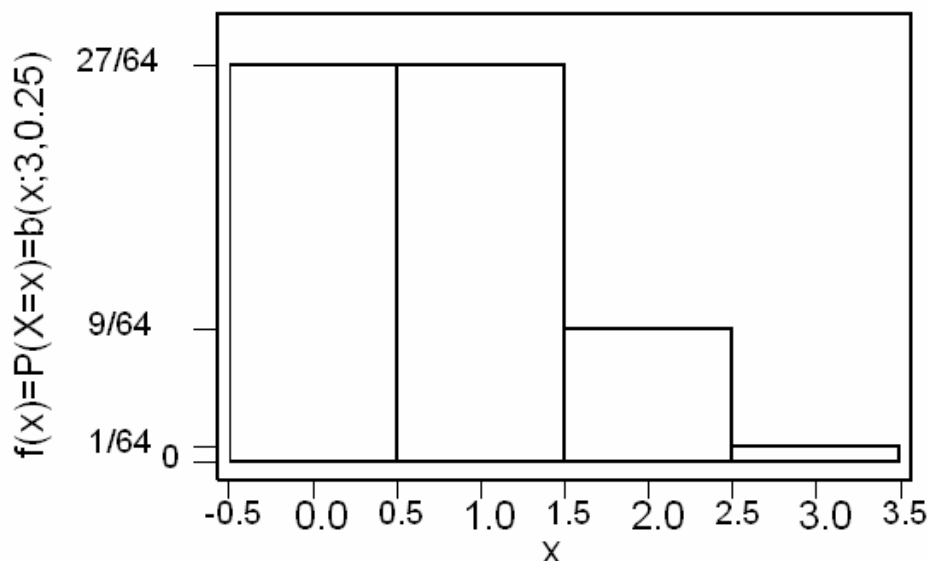
$$f(2) = P(X = 2) = b(2; 3, \frac{1}{4}) = \binom{3}{2} \left(\frac{1}{4}\right)^2 \left(\frac{3}{4}\right)^1 = \frac{9}{64}$$

$$f(3) = P(X = 3) = b(3; 3, \frac{1}{4}) = \binom{3}{3} \left(\frac{1}{4}\right)^3 \left(\frac{3}{4}\right)^0 = \frac{1}{64}$$

The probability distribution of X is

x	$f(x) = P(X=x) = b(x; 3, 1/4)$
0	27/64
1	27/64
2	9/64
3	1/64

$X \sim \text{Binomial}(3, 0.25)$



Theorem 5.2:

The mean and the variance of the binomial distribution $b(x; n, p)$ are:

$$\begin{aligned} \mu &= np \\ \sigma^2 &= np(1-p) \end{aligned}$$

Example:

In the previous example, find the expected value (mean) and the variance of the number of defective items.

Solution:

- X = number of defective items
- We need to find $E(X)=\mu$ and $\text{Var}(X)=\sigma^2$
- We found that $X \sim \text{Binomial}(n,p)=\text{Binomial}(3,1/4)$
- $n=3$ and $p=1/4$

The expected number of defective items is

$$E(X)=\mu = n p = (3) (1/4) = 3/4 = 0.75$$

The variance of the number of defective items is

$$\text{Var}(X)=\sigma^2 = n p (1 - p) = (3) (1/4) (3/4) = 9/16 = 0.5625$$

Example:

In the previous example, find the following probabilities:

- (1) The probability of getting at least two defective items.
- (2) The probability of getting at most two defective items.

Solution:

$X \sim \text{Binomial}(3,1/4)$

$$f(x) = P(X = x) = b(x;3, \frac{1}{4}) = \begin{cases} \binom{3}{x} \left(\frac{1}{4}\right)^x \left(\frac{3}{4}\right)^{3-x} & \text{for } x = 0, 1, 2, 3 \\ 0 & \text{otherwise} \end{cases}$$

.x	.f(x)=P(X=x)=b(x;3,1/4)
0	27/64
1	27/64
2	9/64
3	1/64

- (1) The probability of getting at least two defective items:

$$P(X \geq 2) = P(X=2) + P(X=3) = f(2) + f(3) = \frac{9}{64} + \frac{1}{64} = \frac{10}{64}$$

- (2) The probability of getting at most two defective item:

$$\begin{aligned} P(X \leq 2) &= P(X=0) + P(X=1) + P(X=2) \\ &= f(0) + f(1) + f(2) = \frac{27}{64} + \frac{27}{64} + \frac{9}{64} = \frac{63}{64} \end{aligned}$$

or

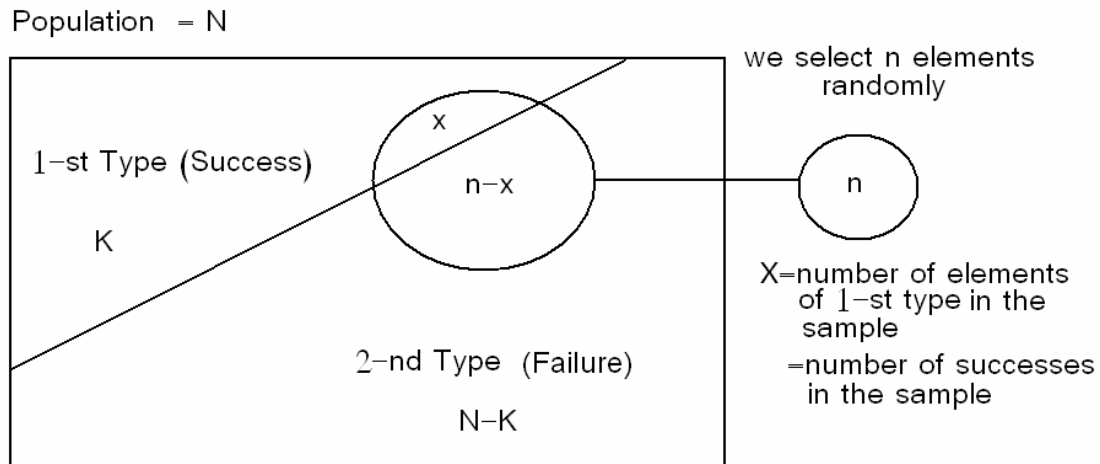
$$P(X \leq 2) = 1 - P(X > 2) = 1 - P(X=3) = 1 - f(3) = 1 - \frac{1}{64} = \frac{63}{64}$$

Example 5.4: Reading assignment

Example 5.5: Reading assignment

Example 5.6: Reading assignment

5.4 Hypergeometric Distribution :



- Suppose there is a population with 2 types of elements:
 - 1-st Type = success
 - 2-nd Type = failure
- N = population size
- K = number of elements of the 1-st type
- $N - K$ = number of elements of the 2-nd type
- We select a sample of n elements at random from the population
- Let X = number of elements of 1-st type (number of successes) in the sample
- We need to find the probability distribution of X .

There are two methods of selection:

1. selection with replacement
2. selection without replacement

(1) If we select the elements of the sample at random and with replacement, then

$$X \sim \text{Binomial}(n, p); \text{ where } p = \frac{K}{N}$$

(2) Now, suppose we select the elements of the sample at random and 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 distribution of X is given by:

$$f(x) = P(X = x) = h(x; N, n, K)$$

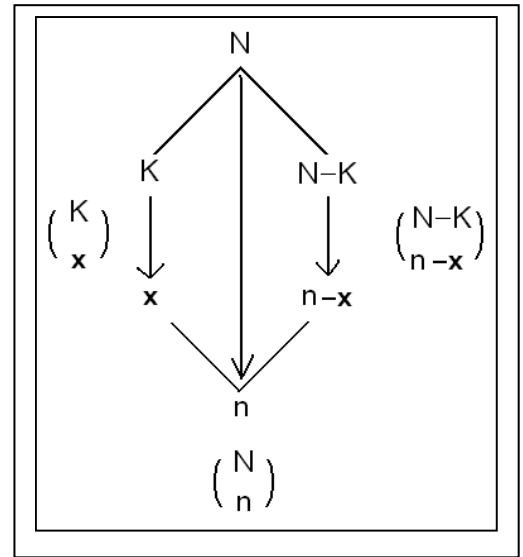
$$= \begin{cases} \frac{\binom{K}{x} \times \binom{N-K}{n-x}}{\binom{N}{n}}; & x = 0, 1, 2, \dots, n \\ 0; & \text{otherwise} \end{cases}$$

Note that the values of X must satisfy:

$$0 \leq x \leq K \quad \text{and} \quad 0 \leq n-x \leq N-K$$

\Leftrightarrow

$$0 \leq x \leq K \quad \text{and} \quad n-N+K \leq x \leq n$$

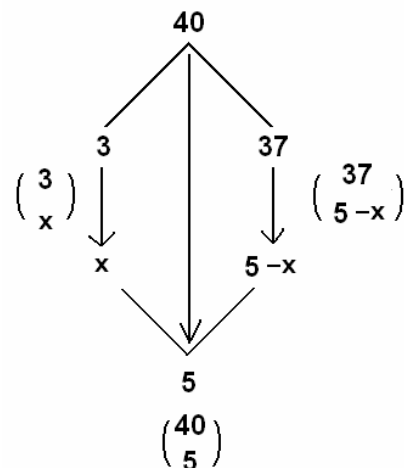
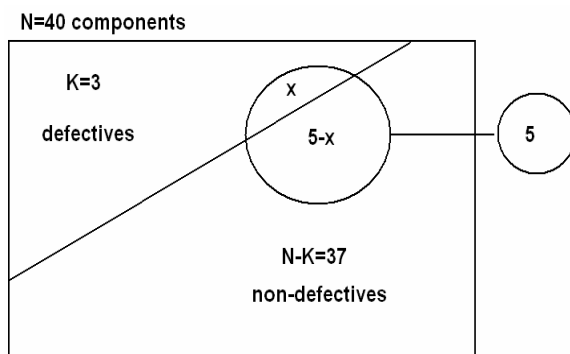


Example 5.8: Reading assignment

Example 5.9:

Lots of 40 components each are called acceptable if they contain no more than 3 defectives. The procedure for sampling the lot is to select 5 components at random (without replacement) and to reject the lot if a defective is found. What is the probability that exactly one defective is found in the sample if there are 3 defectives in the entire lot.

Solution:



- Let X= number of defectives in the sample
- N=40, K=3, and n=5
- X has a hypergeometric distribution with parameters N=40, n=5, and K=3.
- $X \sim h(x; N, n, K) = h(x; 40, 5, 3)$.
- The probability distribution of X is given by:

$$f(x) = P(X = x) = h(x;40,5,3) = \begin{cases} \frac{\binom{3}{x} \times \binom{37}{5-x}}{\binom{40}{5}}; & x = 0, 1, 2, \dots, 5 \\ 0; & \text{otherwise} \end{cases}$$

But the values of X must satisfy:

$$0 \leq x \leq K \quad \text{and} \quad n - N + K \leq x \leq n \Leftrightarrow 0 \leq x \leq 3 \quad \text{and} \quad -32 \leq x \leq 5$$

Therefore, the probability distribution of X is given by:

$$f(x) = P(X = x) = h(x;40,5,3) = \begin{cases} \frac{\binom{3}{x} \times \binom{37}{5-x}}{\binom{40}{5}}; & x = 0, 1, 2, 3 \\ 0; & \text{otherwise} \end{cases}$$

Now, the probability that exactly one defective is found in the sample is

$$.f(1) = P(X=1) = h(1;40,5,3) = \frac{\binom{3}{1} \times \binom{37}{5-1}}{\binom{40}{5}} = \frac{\binom{3}{1} \times \binom{37}{4}}{\binom{40}{5}} = 0.3011$$

Theorem 5.3:

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

$$\mu = n \frac{K}{N}$$

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

Example 5.10:

In Example 5.9, find the expected value (mean) and the variance of the number of defectives in the sample.

Solution:

- X = number of defectives in the sample
- We need to find $E(X) = \mu$ and $\text{Var}(X) = \sigma^2$
- We found that $X \sim h(x;40,5,3)$
- $N=40$, $n=5$, and $K=3$

The expected number of defective items is

$$E(X)=\mu = n \frac{K}{N} = 5 \times \frac{3}{40} = 0.375$$

The variance of the number of defective items is

$$\text{Var}(X)=\sigma^2 = n \frac{K}{N} \left(1 - \frac{K}{N}\right) \frac{N-n}{N-1} = 5 \times \frac{3}{40} \left(1 - \frac{3}{40}\right) \frac{40-5}{40-1} = 0.311298$$

Relationship to the binomial distribution:

* Binomial distribution: $b(x;n,p) = \binom{n}{x} p^x (1-p)^{n-x}$; $x = 0, 1, \dots, n$

* Hypergeometric distribution: $h(x;N,n,K) = \frac{\binom{K}{x} \binom{N-K}{n-x}}{\binom{N}{n}}$; $x = 0, 1, \dots, n$

If n is small compared to N and K , then the hypergeometric distribution $h(x;N,n,K)$ can be approximated by the binomial distribution $b(x;n,p)$, where $p = \frac{K}{N}$; i.e., for large N and K and small n , we have:

$$h(x;N,n,K) \approx b(x;n, \frac{K}{N})$$

$$\frac{\binom{K}{x} \binom{N-K}{n-x}}{\binom{N}{n}} \approx \binom{n}{x} \left(\frac{K}{N}\right)^x \left(1 - \frac{K}{N}\right)^{n-x}; x = 0, 1, \dots, n$$

Note:

If n is small compared to N and K , then there will be almost no difference between selection without replacement and selection with replacement $\left(\frac{K}{N} \approx \frac{K-1}{N-1} \approx \dots \approx \frac{K-n+1}{N-n+1}\right)$.

Example 5.11:

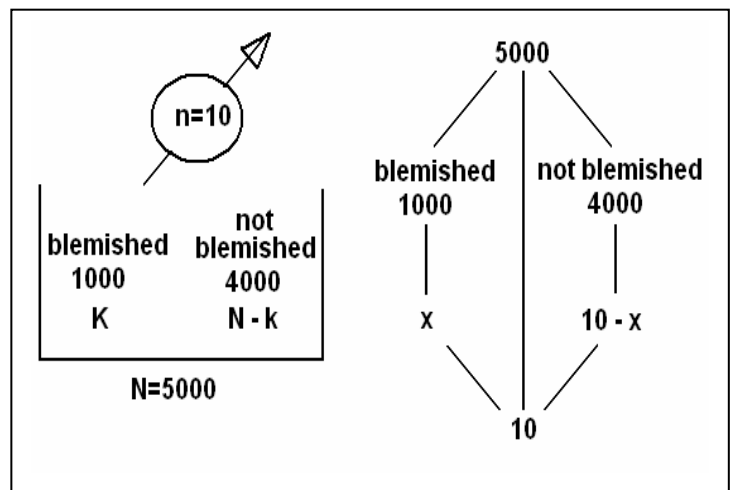
$$N=5000 \quad K=1000 \quad n=10$$

X = Number of blemished tires
in the Sample

$$X \sim h(x; 5000, 10, 1000)$$

The exact probability is

$$\begin{aligned} P(X=3) &= \frac{\binom{1000}{3} \binom{4000}{7}}{\binom{5000}{10}} \\ &= \underline{0.201477715} \\ &\approx \underline{0.201} \end{aligned}$$



Since $n=10$ is small relative to $N=5000$ and $K=4000$, we can approximate the hypergeometric probabilities using binomial probabilities as follows:

$$.n=10 \quad (\text{no. of trials})$$

$$.p=K/N=1000/5000=0.2 \quad (\text{probability of success})$$

$$X \sim h(x; 5000, 10, 1000) \approx b(x; 10, 0.2)$$

$$\begin{aligned} P(X=3) &\approx \binom{10}{3} (0.2)^3 (0.8)^7 = \underline{0.201326592} \\ &\approx \underline{0.201} \end{aligned}$$

5.6 Poisson Distribution:

- Poisson experiment is an experiment yielding numerical values of a random variable that count the number of outcomes occurring in a given time interval or a specified region denoted by t .

X = The number of outcomes occurring in a given time interval or a specified region denoted by t .

- Example:
 1. X = number of field mice per acre ($t= 1$ acre)
 2. X = number of typing errors per page ($t=1$ page)
 3. X =number of telephone calls received every day ($t=1$ day)
 4. X =number of telephone calls received every 5 days ($t=5$ days)
- Let λ be the average (mean) number of outcomes per unit time or unit region ($t=1$).
- The average (mean) number of outcomes (mean of X) in the time interval or region t is:

$$\mu = \lambda t$$

- The random variable X is called a Poisson random variable with parameter μ ($\mu=\lambda t$), and we write $X \sim \text{Poisson}(\mu)$, if its probability distribution is given by:

$$f(x) = P(X = x) = p(x; \mu) = \begin{cases} \frac{e^{-\mu} \mu^x}{x!} & ; \quad x = 0, 1, 2, 3, \dots \\ 0 & ; \quad \textit{otherwise} \end{cases}$$

Theorem 5.5:

The mean and the variance of the Poisson distribution $\text{Poisson}(x; \mu)$ are:

$$\begin{aligned} \mu &= \lambda t \\ \sigma^2 &= \mu = \lambda t \end{aligned}$$

Note:

- λ is the average (mean) of the distribution in the unit time ($t=1$).
- If X =The number of calls received in a month (unit time $t=1$ month) and $X \sim \text{Poisson}(\lambda)$, then:

(i) $Y =$ number of calls received in a year.

$$Y \sim \text{Poisson}(\mu); \quad \mu=12\lambda \quad (t=12)$$

(ii) $W =$ number of calls received in a day.

$$W \sim \text{Poisson}(\mu); \quad \mu=\lambda/30 \quad (t=1/30)$$

Example 5.16: Reading Assignment

Example 5.17: Reading Assignment

Example:

Suppose that the number of typing errors per page has a Poisson distribution with average 6 typing errors.

(1) What is the probability that in a given page:

(i) The number of typing errors will be 7?

(ii) The number of typing errors will be at least 2?

(2) What is the probability that in 2 pages there will be 10 typing errors?

(3) What is the probability that in a half page there will be no typing errors?

Solution:

(1) $X =$ number of typing errors per page.

$$X \sim \text{Poisson}(6) \quad (t=1, \lambda=6, \mu=\lambda t=6)$$

$$f(x) = P(X = x) = p(x;6) = \frac{e^{-6} 6^x}{x!}; \quad x = 0, 1, 2, \dots$$

$$(i) \quad f(7) = P(X = 7) = p(7;6) = \frac{e^{-6} 6^7}{7!} = 0.13768$$

$$(ii) \quad P(X \geq 2) = P(X=2) + P(X=3) + \dots = \sum_{x=2}^{\infty} P(X = x)$$

$$P(X \geq 2) = 1 - P(X < 2) = 1 - [P(X=0) + P(X=1)]$$

$$= 1 - [f(0) + f(1)] = 1 - \left[\frac{e^{-6} 6^0}{0!} + \frac{e^{-6} 6^1}{1!} \right]$$

$$= 1 - [0.00248 + 0.01487]$$

$$= 1 - 0.01735 = 0.982650$$

(2) $X =$ number of typing errors in 2 pages

$$X \sim \text{Poisson}(12) \quad (t=2, \lambda=6, \mu=\lambda t=12)$$

$$f(x) = P(X = x) = p(x;12) = \frac{e^{-12} 12^x}{x!}; \quad x = 0, 1, 2, \dots$$

$$f(10) = P(X = 10) = \frac{e^{-12} 12^{10}}{10!} = 0.1048$$

(3) X = number of typing errors in a half page.

$X \sim \text{Poisson}(3)$ ($t=1/2, \lambda=6, \mu=\lambda t=6/2=3$)

$$f(x) = P(X = x) = p(x;3) = \frac{e^{-3} 3^x}{x!}; \quad x = 0, 1, 2, \dots$$

$$P(X = 0) = \frac{e^{-3}(3)^0}{0!} = 0.0497871$$

Theorem 5.6: (Poisson approximation for binomial distribution):

Let X be a binomial random variable with probability distribution $b(x;n,p)$. If $n \rightarrow \infty$, $p \rightarrow 0$, and $\mu=np$ remains constant, then the binomial distribution $b(x;n,p)$ can be approximated by Poisson distribution $p(x;\mu)$.

- For large n and small p we have:

$$b(x;n,p) \approx \text{Poisson}(\mu) \quad (\mu=np)$$

$$\binom{n}{x} p^x (1-p)^{n-x} \approx \frac{e^{-\mu} \mu^x}{x!}; \quad x = 0, 1, \dots, n; \quad (\mu = np)$$

Example 5.18:

X = number of items producing bubbles in a random sample of 8000 items

$n=8000$ and $p=1/1000 = 0.001$

$X \sim b(x;8000, 0.001)$

The exact probability is:

$$P(X < 7) = P(X \leq 6) = \sum_{x=0}^6 \binom{8000}{x} (0.001)^x (0.999)^{8000-x} = \dots = \underline{0.313252}$$

The approximated probability using Poisson approximation:

$n=8000$ (n is large, i.e., $n \rightarrow \infty$)

$p=0.001$ (p is small, i.e. $p \rightarrow 0$)

$\mu=np = 8000(0.001)=8$

$X \approx \text{Poisson}(8)$

$$f(x) = P(X = x) = p(x;8) = \frac{e^{-8} 8^x}{x!}; \quad x = 0, 1, 2, \dots$$

$$P(X < 7) = P(X \leq 6) = \sum_{x=0}^6 \frac{e^{-8} 8^x}{x!} = e^{-8} \sum_{x=0}^6 \frac{8^x}{x!} = \dots = \underline{0.313374}$$