

Week 7

Theorem: Let a and b be constants and X be either discrete or a continuous random variable.

$$E(aX + b) = aE(X) + b$$

Results:

1. if $a=0$, then $E(b)=b$
2. if $a=1$, then $E(X+b)=E(X)+b$
3. if $b=0$, then $E(a.X)=a.E(X)$
4. if $a=1$ and $b=-E(X)=-\mu$ then $E(X - E(X)) = E(X) - E(E(X)) = \mu - \mu = 0$

Definition: Let X be either a discrete or a continuous random variable. Then it has either a probability function or probability density function. $E(X) = \mu$. The variance of X is defined by

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

Definition: The positive square root of variance is called standard deviation and denoted by σ .

$$\sigma = \sqrt{\sigma^2}$$

Ex: Let X be a discrete random variable. Find $E(X)$, $V(X)$, and σ .

$X=x$	$P(X=x)$	$x.P(X=x)$	$x^2P(X=x)$
0	1/8	0.(1/8)=0	0.(1/8)=0
1	3/8	1.(3/8)=3/8	1.(3/8)=3/8
2	3/8	2.(3/8)=6/8	4.(3/8)=12/8
3	1/8	3.(1/8)=3/8	9.(1/8)=9/8

$$E(X) = 0 + \frac{3}{8} + \frac{6}{8} + \frac{3}{8} = \frac{12}{8}$$

$$E(X^2) = 0 + \frac{3}{8} + \frac{12}{8} + \frac{9}{8} = \frac{24}{8}$$

$$V(X) = E(X^2) - (E(X))^2 = \frac{24}{8} - \left(\frac{12}{8}\right)^2 = \frac{3}{4}$$

$$\sigma = \sqrt{\frac{3}{4}}$$

Ex. Let X be a continuous random variable whose probability density function is defined by

$$f(x) = \begin{cases} 1, & 0 < x < 1 \\ 0, & \text{dd} \end{cases}$$

Find $E(X)$, $V(X)$ and σ

$$E(X) = \int_{-\infty}^{\infty} x \cdot f(x) dx = \int_0^1 x \cdot 1 dx = \frac{x^2}{2} \Big|_0^1 = \frac{1}{2} - 0 = \frac{1}{2}$$

$$E(X^2) = \int_{-\infty}^{\infty} x \cdot f(x) dx = \int_0^1 x^2 \cdot 1 dx = \frac{x^3}{3} \Big|_0^1 = \frac{1}{3} - 0 = \frac{1}{3}$$

$$\sigma^2 = E(X) = E(X^2) - (E(X))^2 = \frac{1}{3} - \left(\frac{1}{2}\right)^2 = \frac{1}{12}$$

$$\sigma = \sqrt{\frac{1}{12}}$$

The Properties of Variance

Theorem: 1. Let a be constant and X be either a discrete or a continuous random variable.

$$V(aX) = a^2 V(X)$$

2. Let b be a constant and X be either a discrete or a continuous random variable.

$$V(X + b) = V(X)$$

Ex: Let X be a discrete random variable and its probability function is defined below.

$X=x$	0	1	2
$P(X=x)$	1/5	2/5	2/5

$V(3X-3)=?$

$$E(X) = 0 + \frac{2}{5} + \frac{4}{5} = \frac{6}{5}$$

$$E(X^2) = 0 + \frac{2}{5} + \frac{8}{5} = \frac{10}{5} = 2$$

$$V(X) = 2 - \frac{36}{25} = \frac{14}{25}$$

$$V(3X - 3) = 9V(X) = 9 \frac{14}{25} = \frac{126}{25}$$

Some Special Discrete Probability Distributions

Definition (Bernoulli Random Variable): Let X be a discrete random variable having two different outcomes, then X is called Bernoulli random variable.

Ex: 1. Toss a coin. (Tail ($T=0$) or Head ($H=1$))

2. A jar contains M black balls and N white balls. A ball is randomly selected from the jar. (Black ball (1) or white ball (0))

3. Defective and undefective tools are stored in a storage. A tool is randomly drawn from the storage. (Defective (1) or Undefective (0))

Definition (Bernoulli Distribution): Let X be Bernoulli random variable either taking 1 or 0. Then its probability function is defined by

$$P(X = 1) = p$$

$$P(X = 0) = 1 - p = q$$

or

$$P(X = x) = p^x(1 - p)^{1-x}, x = 0, 1$$

Theorem: Let X be a Bernoulli random variable. Its expected value and variance are defined by

$$E(X) = p$$

$$V(x) = pq$$

Definition (Binomial random Variable): N independent random Bernoulli experiments are run. Let X denote the number of successes out of n trials. P is a probability denoting the success, $1-p=q$ denotes the probability of failure. Then, X is called a Binomial random variable. The properties of Binomial experiment is defined as follows:

1. Experiment composes of n same trials.
2. Each trial result with 2 different outcomes, which are called success and failure.
3. Each trial has a success probability p and failure probability $1-p=q$.
4. Trials are independent of each other.

Ex: Toss a coin 10 times. Let X show the number of Hs.

2. Suppose that a box contains 8 black and 4 white balls. 3 balls are randomly selected from the box. Let X show the number of black balls.

Theorem (Binomial Distribution) Suppose that n independent Bernoulli trials are executed. For each trial, p denotes the success probability and $1-p$ denotes the failure probability. Let X denote the number of successes out of n trials. Then, binomial distribution of X is defined by

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

Theorem: Let X be a binomial random variable. The expected value and variance of binomial is defined by

$$E(X) = np$$

$$V(x) = npq$$

Ex. Toss a coin 4 times.

a. 2Hs

b. At least one H

c. At most one H

d. $E(x)$ and $V(x)$

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

$$P(X = 2) = \binom{4}{2} \left(\frac{1}{2}\right)^2 \left(1 - \frac{1}{2}\right)^{4-2} = \binom{4}{2} \left(\frac{1}{2}\right)^2 \left(\frac{1}{2}\right)^2$$

b. $P(X \geq 1) = P(X = 1) + P(X = 2) + P(X = 3) + P(X = 4)$

or

$$1 - P(X < 1) = 1 - P(X = 0) = 1 - \binom{4}{0} \left(\frac{1}{2}\right)^0 \left(\frac{1}{2}\right)^4$$

c. $P(X \leq 1) = P(X = 0) + P(X = 1) = \binom{4}{0} \left(\frac{1}{2}\right)^0 \left(\frac{1}{2}\right)^4 + \binom{4}{1} \left(\frac{1}{2}\right)^1 \left(\frac{1}{2}\right)^3$

d. $E(X) = np = 4 \cdot \frac{1}{2} = 2$ ve $V(X) = npq = 4 \cdot \frac{1}{2} \cdot \frac{1}{2} = 1$

