The Mean and Variance Day 4 (1/16/20)

1.9 Some Special Expectations

Definition 1.9.1. The mean of a random variable X is $\mu = E(X)$.

Definition 1.9.2. The variance of a random variable X is $\sigma^2 = E(X - \mu)^2$.

Note that

$$\sigma^{2} = E(X - \mu)^{2} = E(X^{2} - 2\mu X + \mu^{2}) = E(X^{2}) - E(2\mu X) + E(\mu^{2})$$
$$= E(X^{2}) - 2\mu^{2} + \mu^{2} = E(X^{2}) - \mu^{2}$$

which is the easier to compute. The square root of the variance is called the standard deviation.

$$\sigma = \sqrt{E(X-\mu)^2}$$

Theorem 1.9.1. Let a and b be constants. Then

$$Var(aX+b) = a^2 Var(X)$$

Proof.

$$\operatorname{Var}(aX+b) = E\left[(aX+b) - (a\mu+b)\right]^2 = E\left[a(X-\mu)\right]^2 = a^2 E\left(X-\mu\right)^2$$

Ex. Let X have pmf $\frac{x | 2 | 4 | 5 | 6 | 8}{p(x) | \frac{1}{5} | \frac$

1. Find the mean, variance and standard deviation. Solution:

$$\mu = E(X) = \sum xp(x)$$

= $2\left(\frac{1}{5}\right) + 4\left(\frac{1}{5}\right) + 5\left(\frac{1}{5}\right) + 6\left(\frac{1}{5}\right) + 8\left(\frac{1}{5}\right) = 5.0$
 $\sigma^2 = E(X - \mu)^2$
= $(2 - 5)^2\left(\frac{1}{5}\right) + (4 - 5)^2\left(\frac{1}{5}\right) + (5 - 5)^2\left(\frac{1}{5}\right) + (6 - 5)^2\left(\frac{1}{5}\right) + (8 - 5)^2\left(\frac{1}{5}\right)$
= $(9 + 1 + 0 + 1 + 9)\left(\frac{1}{5}\right) = 4.0$

 $\sigma = \sqrt{4.0} = 2.0$

The standard deviation is (an approximation of) "average size of deviation from the mean".

2. Let Y = 3X + 5. Find the mean, variance and standard deviation.

(a) Solution 1: (Find pmf of
$$Y = 3X + 5$$
)
Y has pmf $\frac{y | 11 | 17 | 20 | 23 | 29}{p(y) | \frac{1}{5} | \frac{1$

(b) Solution 2:
$$Var(3X + 5) = (3)^2 Var(X) = (3)^2 4.0 = 6.0$$

Example 1.9.1 Example 1.9.2 Example 1.9.3 (Mean may not exist)

Definition 1.9.3. The moment generating function (mgf) of a random variable X is $M(t) = E(e^{tX})$, for some interval (-h, h) over which the mgf exists.

Example 1.9.4

Theorem 1.9.2. The mgf of X uniquely determines the distribution of X, i.e. if the mgf of X is the same as the mgf of a uniform distribution, then X has a uniform distribution.

Since $M(t) = \int e^{tx} f(x) dx$ then

$$M'(t) = \frac{d}{dt}M(t) = \frac{d}{dt}\int e^{tx}f(x)dx = \int \frac{d}{dt}e^{tx}f(x)dx = \int xe^{tx}f(x)dx$$

Evaluating at t = 0,

$$M'(0) = \int xf(x)dx = E(X)$$

Furthermore,

$$M''(t) = \frac{d}{dt} \int x e^{tx} f(x) dx = \int x^2 e^{tx} f(x) dx$$
$$M''(0) = E(X^2)$$

In general, for any positive integer m,

 $M^{(m)}(0) = E(X^m)$

Example 1.9.5 (mgf may not exist) Example 1.9.7 (mgf of Normal distribution)