Lecture 6

Expected Value and Variance

Manju M. Johny

STAT 330 - Iowa State University

Expected Value

Example 1: Flip a coin 3 times. Let X = # of heads obtained in 3 flips. The probability mass function (pmf) of X is

What number of heads do we "expect" to get?

0 obtained $\frac{1}{8}$ of the time 1 obtained $\frac{3}{8}$ of the time 2 obtained $\frac{3}{8}$ of the time 3 obtained $\frac{1}{8}$ of the time

Intuitively, we can think about taking $0(\frac{1}{8}) + 1(\frac{3}{8}) + 2(\frac{3}{8}) + 3(\frac{1}{8})$ as the "expected" number of heads

Definition

Let X be a discrete random variable. The *expected value* or *expectation* of h(X) is

$$E[h(X)] = \sum_{x} h(x)p_X(x) = \sum_{x} h(x)P(X = x)$$

- The MOST IMPORTANT version of this is when $h(\boldsymbol{x}) = \boldsymbol{x}$

$$E(X) = \sum_{x} x p_X(x) = \sum_{x} x P(X = x)$$

- E(X) is usually denoted by μ
- *E*(*X*) is the weighted average of the *x*'s, where the weights are the probabilities of the *x*'s.

Expected Value Cont.

Example 2: Flip a coin 3 times. Let X = # of heads obtained in 3 flips. The probability mass function (pmf) of X is

Calculate the expected value of X.

$$E(X) = \sum_{x} x p_X(x)$$

= 0P(X = 0) + 1P(X = 1) + 2P(X = 2) + 3P(X = 3)
=

Variance

Variance & Standard Deviation

Definition

The variance of a random variable X is

$$Var(X) = E[(X - E(X))^{2}] = \sum (x - E(X))^{2} \cdot p_{X}(x)$$

The *standard deviation* of a random variable X is

$$\sigma = \sqrt{Var(X)}$$

- Var(X) is usually denoted by σ^2
- Units for variance is squared units of X.
- Units for the standard deviation is same as units for X.

SHORT CUT (use this formula to find variance)

$$Var(X) = E(X^{2}) - [E(X)]^{2}$$
$$= \sum_{x} x^{2} P(X = x) - \left[\sum_{x} x P(X = x)\right]^{2}$$

Example 3: Flip a coin 3 times. Let X = # of heads obtained in 3 flips. The probability mass function (pmf) of X is

Calculate the variance and standard deviation of X.

•
$$E(X) = \sum_{x} x p_X(x) =$$

•
$$E(X^2) = \sum_x x^2 p_X(x) =$$

•
$$Var(X) = E(X^2) - [E(X)]^2 =$$

•
$$\sigma = \sqrt{Var(X)} =$$

Properties of E(X) & Var(X)

Operations with E(X) and Var(X)

X, Y are random variables; a, b, c are constants. Operations with $E(\cdot)$

1.
$$E(aX) = aE(X)$$

2.
$$E(aX + bY + c) = aE(X) + bE(Y) + c$$

Operations with $Var(\cdot)$

3.
$$Var(aX) = a^2 Var(X)$$

4.
$$Var(aX + b) = a^2 Var(X)$$

 Var(aX + bY) = a²Var(X) + b²Var(Y) + 2abCov(X, Y) (when X,Y are independent, Cov(X,Y) = 0. We'll discuss more about independence and define covariance in Lecture 9)

Chebyshev's Inequality

For any positive real number k, and random variable X with variance σ^2 :

$$P(|X - E(X)| \le k\sigma) \ge 1 - \frac{1}{k^2}$$

 bounds the probability that X lies within a certain number of standard deviations from E(X)