

Lecture 9: Expected Value and Variance

Definition - Let X be a discrete random variable with probability distribution $p(x)$. The expected value of X , $E[X]$, is defined by

$$E[X] = \sum_x x p(x). \sim \text{weighted average.}$$

Example:

Return to the sock example

- 5 brown socks

- 3 green socks

- two socks selected at random

$X = \#$ of brown socks

$$p(0) = P(X=0) = \frac{6}{56}$$

$$p(1) = P(X=1) = \frac{30}{56}$$

$$p(2) = P(X=2) = \frac{20}{56}$$

$$\Rightarrow E[X] = 0 \cdot p(0) + 1 \cdot p(1) + 2 \cdot p(2) = \frac{30}{56} + \frac{40}{56} = \frac{70}{56} = \frac{5}{4} = 1.25$$

Definition - If $E[X] = \mu$, the variance of a random variable X is given by

$$\sigma^2 = V[X] = E[(X-\mu)^2] = \sum_x (x-\mu)^2 p(x):$$

$\sigma =$ standard deviation

Example:

Returning to the socks example:

$$V[X] = (-1.25)^2 \frac{6}{56} + (1-1.25)^2 \frac{30}{56} + (2-1.25)^2 \frac{20}{56}$$

$$= .4018$$

$$\Rightarrow \sigma = \sqrt{V[X]} = .63$$

Theorem - $E[ag(x) + bh(x) + c] = aE[g(x)] + bE[h(x)] + c$

proof:

$$\begin{aligned} E[ag(x) + bh(x) + c] &= \sum_x (ag(x) + bh(x) + c)p(x) \\ &= a \sum_x g(x)p(x) + b \sum_x h(x)p(x) + c \sum_x p(x) \\ &= aE[g(x)] + bE[h(x)] + c. \end{aligned}$$

Example:

X = roll of six side die.

$$E[2X^2 + 1] = 2E[X^2] + 1 = 2(1 + 4 + 9 + 16 + 25 + 36) \cdot \frac{1}{6} + 1 = \frac{94}{3} = 31.33.$$

Theorem - $V[X] = E[X^2] - E[X]^2 = E[X^2] - \mu^2$

proof:

$$\begin{aligned} V[X] &= E[(X - E[X])^2] \\ &= E[X^2 - 2XE[X] + E[X]^2] \\ &= E[X^2] - 2E[XE[X]] + E[X]^2 \\ &= E[X^2] - 2E[X]E[X] + E[X]^2 \\ &= E[X^2] - E[X]^2. \end{aligned}$$