

Linearity of Expectation

Defn: If X is a random variable and g is a function $E[g(X)] = \sum_x g(x)p(x)$

Theorem: For any constants a and b ,
 $E[aX+b] = aE[X] + b$.

Proof: $E[aX+b] = \sum_x (ax+b)p(x)$
 $= \sum_x (axp(x) + bp(x)) = \sum_x axp(x) + \sum_x bp(x)$
 $= a \sum_x xp(x) + b \sum_x p(x) = aE[X] + b$

Ex: You pay \$1 to play the following game.

with prob.
 $1/8$ of heads

A ~~fair~~ coin is flipped up to and including the first head and you are paid 12¢ per flip. Do you expect to win or lose money?

Let X be the number of flips until the first head.

Your expected gain is $E[12X - 100] = 12E[X] - 100$
 $= 12 \cdot 8 - 100 = -4$

because $E[X] = \frac{1}{1/8} = 8$

Theorem: Let X and Y be two random variables, possibly dependent. Then $E[X+Y] = E[X] + E[Y]$.

Proof: Let $X(s), Y(s)$ be the values of X, Y for some $s \in \Omega$.

$$E[X+Y] = \sum_{s \in \Omega} (X(s) + Y(s))P(s) = \sum_{s \in \Omega} X(s)P(s) + \sum_{s \in \Omega} Y(s)P(s)$$
$$= E[X] + E[Y]$$

Ex: Let X be the number of heads when a coin with $P(\text{Heads}) = p$ is flipped n times independently.

Let $X_i = \begin{cases} 1, & \text{if } i^{\text{th}} \text{ flip is Heads} \\ 0, & \text{otherwise} \end{cases}$, for $1 \leq i \leq n$

"Indicator random variables": only have values 0 or 1.

$$\begin{aligned} E[X_i] &= 1 \cdot P(X_i = 1) + 0 \cdot P(X_i = 0) \\ &= P(i^{\text{th}} \text{ flip is Heads}) \end{aligned}$$

$$X = \sum_{i=1}^n X_i$$

$$E[X] = E\left[\sum_{i=1}^n X_i\right] = \sum_{i=1}^n E[X_i] = \sum_{i=1}^n p = np$$

Linearity is special. In general,

$$E[XY] \neq E[X]E[Y]$$

$$E[X^2] \neq (E[X])^2$$

$$E[\sqrt{X}] \neq \sqrt{E[X]}$$

$$E(X!) \neq (E[X])!$$

Variance