

Examples on slide pack 6, slide 13.

$$E[X] = \sum_{a=0}^{10} a p_X(a) = \sum_{a=0}^{10} a P(X=a)$$

$$= \sum_{a=0}^{10} a \cdot \frac{1}{2^a} \binom{10}{a} p^a (1-p)^{10-a}$$

$$= 1 \cdot \binom{10}{1} p^1 (1-p)^9 + 2 \binom{10}{2} p^2 (1-p)^8 + \dots$$

Does this have a nice closed form? Yes,
Coming soon.

Ex: Let X be the number of flips up to and including the first head, when a coin with probability p of heads is flipped independently.
(Geometric random variable)

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

$$p_X(2) = P(X=2) = (1-p)p$$

$$p_X(i) = P(X=i) = (1-p)^{i-1} p$$

$$E[X] = \sum_{i=1}^{\infty} i p_X(i) = \sum_{i=1}^{\infty} i (1-p)^{i-1} p = p \sum_{i=1}^{\infty} i (1-p)^{i-1}$$

$$\text{Calculus: } \sum_{i=0}^{\infty} x^i = \frac{1}{1-x} = (1-x)^{-1}, \text{ if } |x| < 1$$

$$\text{Differentiate: } \sum_{i=0}^{\infty} i x^{i-1} = \frac{1}{(1-x)^2}$$

$$x=1-p: E[X] = p \cdot \frac{1}{(1-(1-p))^2} = \boxed{\frac{1}{p}}$$

$$p = \frac{1}{2} \Rightarrow E[X] = 2$$

$$p = \frac{1}{10} \Rightarrow E[X] = 10$$

$$\sum_{i=0}^{\infty} i x^{i-1} = \sum_{i=1}^{\infty} i x^{i-1}$$

Linearity of Expectation

Defn: If X is a r.v., $E[g(X)] = \sum_a g(a)P_X(a)$

Theorem 1: For any constants a and b ,

$$E[aX+b] = aE[X] + b.$$

Proof: $E[aX+b] = \sum_i (ai+b)P_X(i) = \sum_i aiP_X(i) + \sum_i bP_X(i)$

$$= \sum_i (aiP_X(i) + bP_X(i))$$

$$= \sum_i aiP_X(i) + \sum_i bP_X(i)$$

$$= a \sum_i iP_X(i) + b \sum_i P_X(i)$$

$$= aE[X] + b$$

Ex. A casino charges \$1 to play the following game. They flip a coin with probability $1/8$ of heads until it comes up heads and they pay you 12¢ for each flip. Let X be the number of flips. Do you expect to win or lose money?

$$E[12X - 100] = 12E[X] - 100 = 12 \cdot 8 - 100 = -4.$$

Theorem 2: Let X and Y be two random variables, possibly dependent. Then $E[X+Y] = E[X] + E[Y]$.

Linearity is special. In general,

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

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

$$E[\sqrt{X}] \neq \sqrt{E[X]}, \text{ etc.}$$