

Theory of Probability

Oct 12, 2020

Discrete random variable X ,

$$P[X = x_i] = p(x_i)$$

$$\sum_{i=1}^{\infty} p(x_i) = 1$$

Expected value of X :

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

Weighted average of possible values.

If X is a R.V., then so is $Y = g(X)$.

$$\begin{aligned}
 E[Y] &= \sum_{i=1}^{\infty} y_i P[Y = y_i] && \text{if } y_i = g(x_i) \\
 &= \sum_{i=1}^{\infty} g(x_i) P[X = x_i] \\
 &= \sum_{i=1}^{\infty} g(x_i) p(x_i). \\
 &= \underbrace{E[g(X)]}_{Y}
 \end{aligned}$$

Ex: $P[X = -1] = \frac{1}{3}$

Let $Y = X^2$

$$P[X = 0] = \frac{1}{3}$$

Possible values of Y
are 0, 1

$$P[X = 1] = \frac{1}{3}$$

$$P[Y = 0] = P[X = 0] = \frac{1}{3}$$

$$\begin{aligned}
 P[Y = 1] &= P[X = -1 \text{ or } X = 1] \\
 &= \frac{2}{3}.
 \end{aligned}$$

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Variance

- Expected value classifies a RV according to its average value.

$$\text{Var}[X] = E[(X - E[X])^2]$$

$$= E[X^2] - (E[X])^2.$$

$$\text{Var}[aX + b] = a^2 \text{Var}[X].$$

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

is a linear transformation

The n^{th} moment of X is

$$E[X^n] = n^{\text{th}} \text{ moment}.$$

$$\underline{n=1} \quad E[X^n] = E[X] = \text{expected value}$$

$$\underline{n=2} \quad E[X^2] = \text{Var}[X] + (E[X])^2$$

"centered moments"

$$E[\underbrace{(X-\mu)^n}_{X - E[X]}]$$

$$X - E[X]$$

Example: 4.5 from Theoretical Exercises

Let N be a nonnegative integer-valued random variable.

$$P[N = j] = p(j), \quad j \geq 0$$

$$\sum_{j=1}^{\infty} p(j) = 1$$

[2]

(a) For nonnegative a_j , $j \geq 1$, show:

$$\sum_{j=1}^{\infty} (a_1 + \dots + a_j) P[N=j] = \sum_{i=1}^{\infty} a_i P[N \geq i].$$

$$\begin{aligned} E[N] &= \sum_{j=0}^{\infty} j P[N=j] \\ &= \sum_{j=1}^{\infty} j P[N=j]. \end{aligned}$$

Define the function $g(j) = \sum_{i=1}^j a_i$

$$\begin{aligned} E[g(N)] &= \sum_{j=1}^{\infty} g(j) P[N=j] \\ &= a_1 P[N=1] \\ &\quad + (a_1 + a_2) P[N=2] \\ &\quad + (a_1 + a_2 + a_3) P[N=3] \\ &\quad + \dots \\ &= a_1 \sum_{i=1}^{\infty} P[N=i] \\ &\quad + a_2 \sum_{i=2}^{\infty} P[N=i] \\ &\quad + a_3 \sum_{i=3}^{\infty} P[N=i] \end{aligned}$$

$$= \sum_{j=1}^{\infty} a_j \sum_{i=j}^{\infty} P[N=i]$$

$$= \sum_{j=1}^{\infty} a_j P[N \geq j]$$

$$\text{Show that } E[N] = \sum_{i=1}^{\infty} P[N \geq i]$$

Set $a_i = 1$ for all i .

$$\text{Then } E[g(N)] = \sum_{j=1}^{\infty} a_j P[N \geq j]$$

$$= \sum_{j=1}^{\infty} P[N \geq j]$$

and $g(N) = \sum_{i=1}^N a_i = \sum_{i=1}^N 1 = N.$

$$\Rightarrow E[N] = \sum_{j=1}^{\infty} P[N \geq j].$$

And finally, show that : $E[N(N+1)] = 2 \sum_{i=1}^{\infty} i P[N \geq i]$

So first note that

$$\sum_{i=1}^N i = 1 + 2 + 3 + 4 + \dots + N-1 + N$$

$$= \frac{N(N+1)}{2}$$

$$\text{Show that } E\left[\frac{N(N+1)}{2}\right] = \sum_{i=1}^{\infty} i P[N \geq i].$$

$$\text{Set } g(N) = \sum_{i=1}^N i = \frac{N(N+1)}{2}.$$

$$\Rightarrow E[g(N)] = E\left[\frac{N(N+1)}{2}\right] = \sum_{i=1}^{\infty} i P[N \geq i]$$