

Cumulative Distribution Function

For each of the following functions let's determine the values of k for which p is a probability mass function for a discrete variable.

(i) $p(x) = kx$, $x=1,2,\dots,n$

(ii) $p(x) = kx^2$, $x=1,2,\dots,n$

(i) $p(x)$ is a probability function iff

$$\begin{aligned} 1 &= \sum_{x_i} p(x_i) \\ &= \sum_{x=1}^n p(x) \\ &= \sum_{x=1}^n kx \\ &= k \sum_{x=1}^n x \\ 1 &= k \left(\frac{n(n+1)}{2} \right) \\ k &= \frac{2}{n(n+1)} \end{aligned}$$

$p(x)$ is a probability function iff the sum of the probabilities from 1 to n is equal to 1. The sum of all numbers from 1 to n is $\frac{n(n+1)}{2}$ so for a probability function $k = \frac{2}{n(n+1)}$.

(ii)

$$\begin{aligned} 1 &= \sum_{x_i} p(x_i) \\ &= \sum_{x=1}^n kx^2 \\ &= k \sum_{x=1}^n x^2 \\ &= k \left(\frac{n(n+1)(2n+1)}{6} \right) \\ k &= \frac{6}{n(n+1)(2n+1)} \end{aligned}$$

Expected Value

definition - X discrete random variable with values $\{x_1, x_2, x_3, \dots\}$ and function p . Then the expected value (mathematical expectation) is given by

$$E[X] = \sum_{x_i} p(x_i)$$

example - Toss a fair coin once. Let X be 0 if it comes out heads, 1 if tails. Calculate $E[X]$.

What are the possible values of X ? $X = \{0, 1\}$

$$P(X = 0) = P(X = 1) = \frac{1}{2} \Rightarrow P(0) = P(1) = \frac{1}{2}$$

$$\begin{aligned} E[X] &= \sum_{x_i} x_i p(x_i) \\ &= 0 \cdot \frac{1}{2} + 1 \cdot \frac{1}{2} = \frac{1}{2} \end{aligned}$$

example - Suppose that we place the numbers 3, 5, 7, 9, 11, & 13 on otherwise identical tickets. The tickets are placed in a hat and one ticket is selected at random. Let X be the number of the ticket drawn. Find $E[X]$.

$$\begin{aligned} E[X] &= \sum_{x_i} p(x_i) \\ &= 3 \frac{1}{6} + 5 \frac{1}{6} + 7 \frac{1}{6} + 9 \frac{1}{6} + 11 \frac{1}{6} + 13 \frac{1}{6} \\ &= \frac{1}{6} (3 + 5 + 7 + 9 + 11 + 13) \\ &= \frac{48}{6} = 8 \end{aligned}$$

It can be considered a weighted average, although in this case it is evenly weighted.

example - Six numbers are selected at random from the positive integers 1-49 for a winning number.

- Match all 6 numbers and win a grand prize of 1,200,000 dollars.
- Match 5 numbers and win a second prize of 800 dollars.
- Match 5 number and win a third prize of 35 dollars.

Find the expected number of dollars that a player wins in a single play.

Let X be the dollar amount won by the player in a single play. Possible values of $X = \{0, 35, 800, 1.2\text{mil}\}$

$$P(1200000) = P(X = 1200000) = \frac{1}{\binom{49}{6}} \approx 0.00000072$$

$$P(800) = P(X = 800) = \frac{\binom{6}{5} \binom{43}{1}}{\binom{49}{6}} \approx 0.000018$$

$$P(35) = P(X = 35) = \frac{\binom{6}{4} \binom{43}{2}}{\binom{49}{6}} \approx 0.00097$$

$$P(0) = P(X = 0) = 1 - (0.00000072 + 0.000018 + 0.00097) = 0.999011928$$

$$E[X] = 1200000(0.00000072) + 800(0.000018) + 35(0.00097) + 0 \approx 0.13$$

Theorem 3.1. Let X be a discrete random variable with possible values $\{x_1, x_2, x_3, \dots\}$ and probability function p . Let g be a real valued function whose domain contains $\{x_1, x_2, x_3, \dots\}$. Then $g(X)$ is a discrete random variable and

$$E[g(X)] = \sum_{x_i} g(x_i)p(x_i)$$

example - The probability function p of a discrete random variable X is given below

$$p(x) = \begin{cases} \frac{x}{15}, & x = 1, 2, 3, 4, 5 \\ 0, & \text{otherwise} \end{cases}$$

Calculate the expected value of $X(6-X)$. Let $g(x) = x(6-x)$.

$$\begin{aligned} E[g(X)] &= \sum_{x_i} g(x_i)p(x_i) \\ &= \sum_{x=1}^5 x(6-x) \frac{x}{15} \\ &= 5 \cdot \frac{1}{15} + 8 \cdot \frac{2}{15} + 9 \cdot \frac{3}{15} + 8 \cdot \frac{4}{15} + 5 \cdot \frac{5}{15} \\ &= \frac{1}{15}(5 + 6 + 27 + 32 + 25) = \frac{105}{15} = \frac{21}{3} = 7 \end{aligned}$$

definition - Let X be a discrete random variable with possible values $\{x_1, x_2, \dots\}$, probability function p , an expected value $E[X]$. Then the variance of X is defined by

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

and the standard deviation of X

$$\sigma_X = \sqrt{\text{var}[X]}$$

remark - common in the literature to make the substitution $\mu = E[X]$.

$$\begin{aligned}\text{var}[X] &= E[(X - \mu)^2] \\ \sigma_X &= \sqrt{E[(X - \mu)^2]}\end{aligned}$$

example Calculate the variance of the random variable X , where X denotes the number of spots obtained in a single roll of a fair die. $\{1, 2, 3, 4, 5, 6\}$ = possible values of X

$$p(x) = \begin{cases} \frac{1}{6}, & x = 1, 2, 3, 4, 5, 6 \\ 0, & \text{otherwise} \end{cases}$$

$$E[X] = \frac{1}{6}(1 + 2 + 3 + 4 + 5 + 6) = \frac{21}{6} = \frac{7}{2}$$

$$\begin{aligned}\text{var}[X] &= E[(X - E[X])^2] \\ &= E[(X - \frac{7}{2})^2] \\ &= \sum_{x_i} (x_i - \frac{7}{2})^2 p(x_i) = \frac{1}{6} \left[2 \frac{25}{4} + 2 \frac{9}{4} + 2 \frac{1}{4} \right] \\ &= \frac{1}{12} [25 + 9 + 1] = \frac{35}{12}\end{aligned}$$

Theorem 3.1. *Let X be a discrete random variable and a, b are constants. Then*

- (i) $E[aX + b] = aE[X] + b$
- (ii) $\text{var}[aX + b] = a^2 \text{var}[X]$
- (iii) $\sigma_{aX + b} = |a| \sigma_X$

Proof.

$$\begin{aligned} E[aX + b] &= \sum_{x_i} (ax_i + b)p(x_i) \\ &= \sum_{x_i} ax_i p(x_i) + \sum_{x_i} bp(x_i) \\ &= a \sum_{x_i} x_i p(x_i) + b \sum_{x_i} p(x_i) \\ &= aE[X] + b \end{aligned}$$

□

HOMEWORK - 3.4, 3.6, 3.7, 3.10*