

Variance

An important function of X is the “squared deviation” function $g(X) = (X - E(X))^2$. The expected value of g is called the variance $V(X)$ of X and represent the variation of X -values:

$$V(X) = E(X(X - E(X))^2) = E((X - \mu)^2)$$

We can compute variances using the shortcut formula $V(X) = E(X^2) - (E(X))^2$. (Note: $E(X) = \sum x^2 \cdot p(x)$)

The standard deviation of X is the square root of the variance: $SD(X) = \sigma(X) = \sqrt{V(X)}$

Example

Let X be a discrete random variable with mass function defined by $p(1) = 0.45$, $p(3) = 0.25$, and $p(5) = 0.30$. What is the variance of X ?

$$\begin{aligned} E(X^2) &= 1^2 \cdot 0.45 + 3^2 \cdot 0.25 + 5^2 \cdot 0.30 \\ &= 0.45 + 9 \cdot 0.25 + 25 \cdot 0.30 \\ &= 0.45 + 2.25 + 7.50 \\ &= 10.20 \end{aligned}$$

$$\begin{aligned} (E(X))^2 &= (1 \cdot 0.45 + 3 \cdot 0.25 + 5 \cdot 0.30)^2 \\ &= (0.45 + 0.75 + 1.50)^2 \\ &= 2.70^2 \\ &= 7.29 \end{aligned}$$

$$\begin{aligned} V(X) &= E(X^2) - (E(X))^2 \\ &= 10.20 - 7.29 \\ &= 2.91 \end{aligned}$$

Example

Suppose that a coin is unfairly loaded to land on head with probability α and let X be the number of coin tosses required to observe a head. From previous lectures, we know that the pmf p of X is defined by:

$$p(x) = \begin{cases} \alpha(1-\alpha)^{x-1} & x = 1, 2, 3, \dots \\ 0 & \text{otherwise} \end{cases} \quad \text{with } E(X) = \frac{1}{\alpha}$$

What is the variance of X ?

$$E(X) = \frac{1}{\alpha} \Rightarrow (E(X))^2 = \frac{1}{\alpha^2}$$

$$\begin{aligned} E(X^2) &= \sum_{x=1}^{\infty} x^2 \cdot p(x) \\ &= \sum_{x=1}^{\infty} x^2 \cdot \alpha \cdot (1-\alpha)^{x-1} \quad \text{let } q = 1 - \alpha \\ &= \sum_{x=1}^{\infty} x^2 \cdot \alpha \cdot q^{x-1} \\ &= \alpha \sum_{x=1}^{\infty} x^2 \cdot q^{x-1} \quad \text{let } n = x - 1 \Rightarrow x = n + 1 \\ &= \alpha \sum_{n=0}^{\infty} (n+1)^2 \cdot q^n \\ &= \alpha \sum_{n=0}^{\infty} (n^2 + 2n + 1) \cdot q^n \\ &= \alpha \left(\sum_{n=0}^{\infty} n^2 \cdot q^n + 2 \sum_{n=0}^{\infty} n \cdot q^n + \sum_{n=0}^{\infty} q^n \right) \\ &= \alpha \left(\frac{(1-\alpha)(2-\alpha)}{\alpha^3} + \frac{2(1-\alpha)}{\alpha^2} + \frac{1}{\alpha} \right) \\ &= \frac{(1-\alpha)(2-\alpha)}{\alpha^2} + \frac{2(1-\alpha)}{\alpha} + 1 \end{aligned}$$

$$\begin{aligned} V(X) &= E(X^2) - (E(X))^2 \\ &= \frac{(1-\alpha)(2-\alpha)}{\alpha^2} + \frac{2(1-\alpha)}{\alpha} + 1 - \frac{1}{\alpha^2} \\ &= \frac{1-\alpha}{\alpha^2} \end{aligned}$$

If X is a discrete random variable and a and b are any constants,

$$V(aX + b) = a^2 \cdot V(X)$$

Example

Let X be discrete random variable with mass function defined by $p(1) = 0.45$, $p(3) = 0.25$, and $p(5) = 0.30$. Let $g(X) = 3X + 2$ be a function of X . What is the variance of $g(X)$?

$$\begin{aligned} V(3X + 2) &= 3^2 \cdot V(X) \\ &= 9 \cdot V(X) \\ &= 9 \cdot 2.91 \\ &= 26.19 \end{aligned}$$

Example

Suppose that two players $P1$ and $P2$ play a series of games that ends when one of them wins twice. Suppose that each game is played independently and $P1$ wins a game with probability α . Let X be the number of games played in the series. Compute $E(X)$ and $V(X)$. What value of α maximizes $E(X)$?

$$\begin{aligned} \Pr(X = 2) &= \alpha^2 + (1 - \alpha)^2 \\ \Pr(X = 3) &= 2\alpha^2(1 - \alpha) + 2\alpha(1 - \alpha)^2 = 2\alpha(1 - \alpha) \\ E(X) &= 2 \cdot (\alpha^2 + (1 - \alpha)^2) + 3 \cdot (2\alpha(1 - \alpha)) \\ &= -2\alpha^2 + 2\alpha + 2 \end{aligned}$$

$$\begin{aligned} (E(X))^2 &= (-2\alpha^2 + 2\alpha + 2)^2 \\ &= 4\alpha^4 - 8\alpha^3 - 4\alpha^2 + 8\alpha + 4 \end{aligned}$$

$$\begin{aligned} E(X^2) &= 2^2 \cdot (\alpha^2 + (1 - \alpha)^2) + 3^2 \cdot (2\alpha(1 - \alpha)) \\ &= 4 \cdot (\alpha^2 + (1 - \alpha)^2) + 9 \cdot (2\alpha(1 - \alpha)) \\ &= -10\alpha^2 + 10\alpha + 4 \end{aligned}$$

$$\begin{aligned} V(X) &= E(X^2) - (E(X))^2 \\ &= (-10\alpha^2 + 10\alpha + 4) - (4\alpha^4 - 8\alpha^3 - 4\alpha^2 + 8\alpha + 4) \\ &= -4\alpha^4 + 8\alpha^3 - 6\alpha^2 + 2\alpha \end{aligned}$$

In order to maximize $E(X)$, the vertex of $\alpha = -\frac{2}{-4} = \frac{1}{2}$. $a = -2 < 0$, so the graph is concave down, so the POI is a maximum. Therefore $\alpha = \frac{1}{2}$ maximizes $E(X)$.