

Recall:

$X$ : Discrete Random Variable

$p(x)$ : Probability distribution

$$\text{Expected Value of } X = E(X) = \sum_{x: p(x) \neq 0} x p(x)$$

$$\text{Variance of } X = \text{Var}(X) = \sum_{x: p(x) \neq 0} (x - \mu)^2 p(x)$$

Another Interpretation of  $\text{Var}(X)$

$$E(X) = \sum_{x: p(x) \neq 0} x p(x)$$

probability of getting  $x$

value assigned =  $x$

$$\text{Var}(X) = \sum_{x: p(x) \neq 0} (x - \mu)^2 p(x)$$

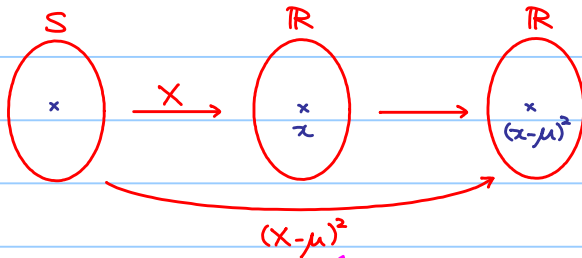
probability of getting  $x$

value assigned =  $(x - \mu)^2$

$$\text{Var}(X) = \sum_{x: p(x) \neq 0} (x - \mu)^2 p(x)$$

probability of getting  $x$

value assigned =  $(x - \mu)^2$



$(x - \mu)^2$

regarded as another random variable

$$\therefore \text{Var}(X) = E((X - \mu)^2)$$

Alternative Formula for  $\text{Var}(X)$

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

$$= \sum_{x:p(x) \neq 0} (x-\mu)^2 p(x)$$

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

$$= \sum_{x:p(x) \neq 0} (x-\mu)^2 p(x)$$

$$= \sum_{x:p(x) \neq 0} (x^2 - 2x\mu + \mu^2) p(x)$$

$$= \sum_{x:p(x) \neq 0} x^2 p(x) - \sum_{x:p(x) \neq 0} 2x\mu p(x) + \sum_{x:p(x) \neq 0} \mu^2 p(x)$$

$$= \sum_{x:p(x) \neq 0} x^2 p(x) - 2\mu \sum_{x:p(x) \neq 0} x p(x) + \mu^2 \sum_{x:p(x) \neq 0} p(x)$$

$$= E(X^2) - 2\mu \cdot \mu + \mu^2 \cdot 1$$

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

$$= E(X^2) - E(X)^2$$

e.g. Roll a die

Let  $X$  be the random variable denotes the number facing up.

$$E(X) = \frac{7}{2} = 3.5$$

$$\begin{aligned} \text{Var}(X) &= (1-3.5)^2 \cdot \frac{1}{6} + (2-3.5)^2 \cdot \frac{1}{6} + (3-3.5)^2 \cdot \frac{1}{6} + \\ &\quad (4-3.5)^2 \cdot \frac{1}{6} + (5-3.5)^2 \cdot \frac{1}{6} + (6-3.5)^2 \cdot \frac{1}{6} \\ &= \frac{35}{12} \end{aligned}$$

(Verification of the alternative formula)

$$\begin{aligned} E(X^2) &= 1^2 \cdot \frac{1}{6} + 2^2 \cdot \frac{1}{6} + 3^2 \cdot \frac{1}{6} + 4^2 \cdot \frac{1}{6} + 5^2 \cdot \frac{1}{6} + 6^2 \cdot \frac{1}{6} \\ &= \frac{91}{6} \end{aligned}$$

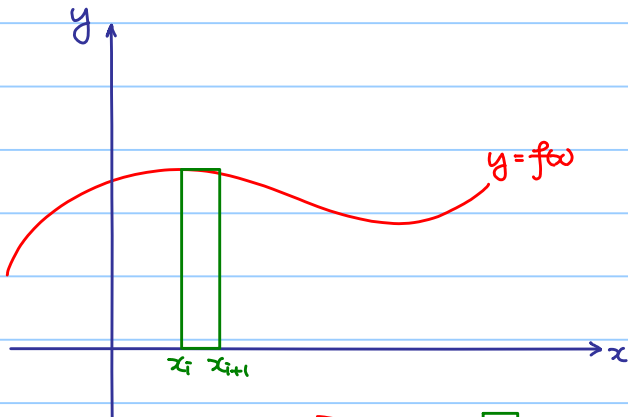
$$E(X)^2 = \left(\frac{7}{2}\right)^2 = \frac{49}{4}$$

$$\text{Var}(X) = E(X^2) - E(X)^2 = \frac{91}{6} - \frac{49}{4} = \frac{35}{12}$$

## Expected Value and Variance of Continuous Random Variables

$X$ : Continuous Random Variable

$f(x)$ : Probability distribution



$$P(x_i \leq X \leq x_{i+1}) = \int_{x_i}^{x_{i+1}} f(x) dx \approx f(x_i) \Delta x$$

The diagram shows two green rectangles. The first rectangle is under the curve between  $x_i$  and  $x_{i+1}$ . The second rectangle is a simpler approximation with the same width  $\Delta x$  and height  $f(x_i)$ . A double-headed arrow below the second rectangle indicates its width is  $\Delta x$ .

$$E(X) = \lim_{n \rightarrow \infty} \sum_{i=1}^n \underset{\substack{\uparrow \\ \text{value}}}{x_i} \cdot \underset{\substack{\uparrow \\ \text{probability}}}{f(x_i) \Delta x} = \int_{-\infty}^{+\infty} x f(x) dx$$

$$\therefore \mu = E(X) = \int_{-\infty}^{+\infty} x f(x) dx$$

Similarly,

$$\text{Var}(X) = \lim_{n \rightarrow \infty} \sum_{i=1}^n \underset{\substack{\uparrow \\ \text{value}}}{(x_i - \mu)^2} \cdot \underset{\substack{\uparrow \\ \text{probability}}}{f(x_i) \Delta x} = \int_{-\infty}^{+\infty} (x - \mu)^2 f(x) dx$$

$$\therefore \text{Var}(X) = E((X - \mu)^2) = \int_{-\infty}^{+\infty} (x - \mu)^2 f(x) dx$$



Alternative Formula for  $\text{Var}(X)$  :

$$\text{Var}(X) = \int_{-\infty}^{+\infty} (x - \mu)^2 f(x) dx$$

$$= \int_{-\infty}^{+\infty} (x^2 - 2x\mu + \mu^2) f(x) dx$$

$$\text{Var}(X) = \int_{-\infty}^{+\infty} (x-\mu)^2 f(x) dx$$

$$= \int_{-\infty}^{+\infty} (x^2 - 2x\mu + \mu^2) f(x) dx$$

$$= \int_{-\infty}^{+\infty} x^2 f(x) dx - 2\mu \int_{-\infty}^{+\infty} x f(x) dx + \mu^2 \int_{-\infty}^{+\infty} f(x) dx$$

$$= E(X^2) - 2\mu \cdot \mu + \mu^2 \cdot 1$$

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

$$= E(X^2) - E(X)^2$$

## Uniform Density Function

$$f(x) = \begin{cases} \frac{1}{b-a} & \text{if } a \leq x \leq b \\ 0 & \text{otherwise} \end{cases}$$

and let  $X$  be random variable whose probability distribution is  $f(x)$ .

$$\mu = E(X) = \int_{-\infty}^{+\infty} x f(x) dx$$

$$= \int_a^b \frac{x}{b-a} dx$$

$$= \left[ \frac{x^2}{2(b-a)} \right]_a^b$$

$$= \frac{a+b}{2}$$

$$E(X^2) = \int_{-\infty}^{+\infty} x^2 f(x) dx$$

$$= \int_a^b \frac{x^2}{b-a} dx$$

$$= \left[ \frac{x^3}{3(b-a)} \right]_a^b$$

$$= \frac{b^2+ab+a^2}{3}$$

$$\text{Var}(X) = E(X^2) - E(X)^2 = \frac{b^2+ab+a^2}{3} - \left(\frac{a+b}{2}\right)^2 = \frac{(b-a)^2}{12}$$

## Exponential Density Function

$$\text{Let } f(x) = \begin{cases} \lambda e^{-\lambda x} & \text{if } x \geq 0 \\ 0 & \text{if } x < 0 \end{cases}, \text{ where } \lambda > 0$$

and let  $X$  be random variable whose probability distribution is  $f(x)$ .

$$E(x) = \int_{-\infty}^{+\infty} x f(x) dx$$

$$= \int_0^{+\infty} \lambda x e^{-\lambda x} dx$$

↑ why?

$$= \int_0^{+\infty} x d(-e^{-\lambda x})$$

$$= \underbrace{[-x e^{-\lambda x}]_0^{+\infty}}_0 - \int_0^{+\infty} -e^{-\lambda x} dx$$

$$= \int_0^{+\infty} e^{-\lambda x} dx$$

$$= \left[-\frac{1}{\lambda} e^{-\lambda x}\right]_0^{+\infty}$$

$$= \frac{1}{\lambda}$$

$$E(X^2) = \int_{-\infty}^{+\infty} x^2 f(x) dx = \int_0^{+\infty} \lambda x^2 e^{-\lambda x} dx = \dots = \frac{2}{\lambda^2}$$

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

$$\text{Var}(X) = E(X^2) - E(X)^2 = \frac{1}{\lambda^2}$$