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Note Title

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If $X \sim \text{Poisson}(\lambda)$, what is $E[X]$

$$p_X(i) = \frac{e^{-\lambda} \lambda^i}{i!} \quad (\text{pdf of } X \sim \text{Poisson}(\lambda)) \quad (0 \leq i < \infty)$$

$$E[X] = \sum_{i=0}^{\infty} i p_X(i) = \sum_{i=0}^{\infty} i \frac{e^{-\lambda} \lambda^i}{i!} = \sum_{i=1}^{\infty} i \frac{e^{-\lambda} \lambda^i}{i!}$$

$$= e^{-\lambda} \sum_{i=1}^{\infty} i \frac{\lambda^i}{i!} = \lambda e^{-\lambda} \sum_{i=1}^{\infty} \frac{\lambda^{i-1}}{i(i-1)(i-2)\dots 1} = \lambda e^{-\lambda} \sum_{i=1}^{\infty} \frac{\lambda^{i-1}}{(i-1)!}$$

$$= \lambda e^{-\lambda} \sum_{k=0}^{\infty} \frac{\lambda^k}{k!} = \lambda e^{-\lambda} e^{\lambda} = \lambda$$

Moment of an r.v. The i -th moment of r.v. X , denoted $E[X^i]$ is

$$E[X^i] = \sum_x x^i p_X(x) \quad (\text{discrete case}) \quad x \quad X$$

$$E[X^i] = \int_{-\infty}^{\infty} x^i f_X(x) dx \quad (\text{continuous case})$$

More generally, the expectation of function $g(\cdot)$ of the r.v. X is:

$$E[g(X)] = \sum_x g(x) p_X(x) \quad (\text{discrete case})$$

$$E[g(X)] = \int_{-\infty}^{\infty} g(x) f_X(x) dx$$

Example (p. 46 [H])

$$X = \begin{cases} 0 & \omega / \text{prob. } 0.2 \\ 1 & \omega / \text{prob. } 0.5 \\ 2 & \omega / \text{prob. } 0.3 \end{cases}$$

$$\begin{aligned} E[2X^2 + 3] &= \sum_x (2x^2 + 3) p_X(x) = (2 \cdot 0^2 + 3) p_X(0) + (2 \cdot 1^2 + 3) p_X(1) + \\ &+ (2 \cdot 2^2 + 3) p_X(2) = 3 \times 0.2 + 5 \times 0.5 + 11 \times 0.3 = \\ &= 0.6 + 2.5 + 3.3 = 6.4 \end{aligned}$$

Defn. 3.17 Variance

The square of how much we expect X to differ from its mean, $E[X]$.

$$\text{Var}(X) = E[(X - E[X])^2] = \sum_x (x - E[X])^2 p_X(x) \quad (\text{discrete case})$$

$$= \int_{-\infty}^{\infty} (x - E[X])^2 f_X(x) dx \quad (\text{continuous case})$$

If $X \sim \text{Bernoulli}(p)$, what is $\text{Var}(X)$?

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

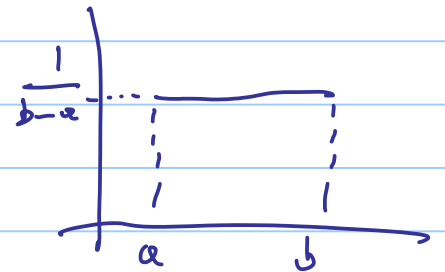
$$= E[(X - p)^2] = \sum_i (i - p)^2 p_X(i) = (0 - p)^2 p_X(0) + (1 - p)^2 p_X(1) =$$

$$= \underbrace{(0 - p)^2}_{(1-p)} + \underbrace{(1 - p)^2}_p = (1 - p)[p^2 + (1 - p)p] = (1 - p)(p^2 + p - p^2)$$

$$= p - p^2 = p(1 - p)$$

If $X \sim \text{Uniform}(a, b)$, what is $\text{Var}(X)$?

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



$$E[X] = \int_a^b x \frac{1}{b-a} dx = \frac{1}{b-a} \int_a^b x dx = \frac{1}{b-a} \left. \frac{x^2}{2} \right|_a^b = \frac{1}{b-a} \left(\frac{b^2 - a^2}{2} \right)$$

$$= \frac{1}{\cancel{b-a}} \frac{\cancel{(b-a)}(b+a)}{2} = \frac{b+a}{2} = \frac{a+b}{2}$$

$$(*) E \left[\left(X - \frac{a+b}{2} \right)^2 \right] = \int_a^b \left(x - \frac{a+b}{2} \right)^2 \frac{1}{b-a} dx = \dots = \frac{(b-a)^2}{12}$$

Joint probability & independence (3.10 [H])

The joint probability mass function between discrete r.v.s X and Y is

$$p_{XY}(x, y) = \mathbb{P} \left\{ \underbrace{X=x, Y=y}_{\text{an event}} \right\}$$

The joint probability density function.....

$$\int_c^d \int_a^b f_{XY}^{(x,y)} dx dy = \mathbb{P} \{ a < X < b, c < Y < d \}$$