

Probability Functions

PMF/PDF, CDF, Expected Value, Variance

General

(*independent assumed)

$$\mathbb{E}[X] = \mu = \sum_x xp_X(x) = \int_{-\infty}^{\infty} xf_X(x)dx$$

$$\mathbb{E}[\alpha X + \beta Y] = \alpha \mathbb{E}[X] + \beta \mathbb{E}[Y]$$

$$*\mathbb{E}[XY] = \mathbb{E}[X]\mathbb{E}[Y]$$

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

$$*\text{Var}[X + Y] = \text{Var}[X] + \text{Var}[Y]$$

$$\sigma_X := \sqrt{\text{Var}[X]} = \sqrt{\mathbb{E}[(X - \mu)^2]}$$

$X \sim \text{Bernoulli}(p)$

$$p_X(x) = \begin{cases} 1 - p & \text{if } x = 0 \\ p & \text{if } x = 1 \end{cases}$$

$$\mathbb{E}[X] = p \quad \text{Var}[X] = p(1 - p)$$

$X \sim \text{Binomial}(n, p)$

$$p_X(k) = \mathbb{P}(X = k) = \binom{n}{k} p^k (1 - p)^{n - k}$$

$$\mathbb{E}[X] = np \quad \text{Var}[X] = np(1 - p)$$

$X \sim \text{Geometric}(p)$

$$p_X(k) = \mathbb{P}(X = k) = p(1 - p)^{k - 1}$$

$$\mathbb{E}[X] = \frac{1}{p} \quad \text{Var}[X] = \frac{1 - p}{p^2}$$

$X \sim \text{Poisson}(\lambda)$

$$p(k) = e^{-\lambda} \frac{\lambda^k}{k!}, \quad k = 0, 1, 2, 3$$

$$\mathbb{E}[X] = \lambda \quad \text{Var}[X] = \lambda$$

$X \sim \text{Uniform}(a, b)$

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

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

$$\mathbb{E}[X] = \frac{a + b}{2} \quad \text{Var}[X] = \frac{(b - a)^2}{12}$$

$X \sim \text{Exponential}(\lambda)$

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

$$F_X(x) = 1 - e^{-\lambda x}$$

$$\mathbb{E}[X] = \frac{1}{\lambda} \quad \text{Var}[X] = \frac{1}{\lambda^2}$$

Joint PMF

$$p(x, y) = \mathbb{P}(X = x, Y = y)$$

Normal Distribution

$X \sim \mathcal{N}(\mu, \sigma^2)$

$$f_X(x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x - \mu)^2}{2\sigma^2}}$$

$$\mathbb{E}[X] = \mu \quad \text{Var}[X] = \sigma^2$$

Standard Normal Distribution

$\mathcal{N}(0, 1)$

$$f_X(x) := \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}$$

$$\mathbb{E}[Z] = 0 \quad \text{Var}[Z] = 1.$$

Convert:

$$Z = \frac{X - \mu}{\sigma}$$

Central Limit Theorem

$$Z_n = \frac{S_n - n\mu}{\sigma\sqrt{n}}, \quad S_n = X_1 + \dots + X_n$$

$$\lim_{n \rightarrow \infty} \mathbb{P}(Z_n \leq z) = \Phi(z)$$

Bayesian Statistic