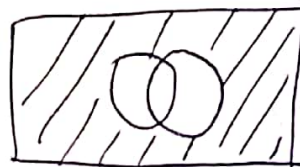


⇒

$$\overline{A \cup B} = A' \cap B'$$



$$1 - P(A \cup B) = P(A' \cap B')$$

$$1 - (P(A) + P(B) - P(A \cap B)) = P(A' \cap B')$$

$$P(\bar{A})P(\bar{B}) = P(\bar{A} \cap \bar{B})$$

⇒

Continuous Probability Function

$$f(x)$$

$$\int_{-\infty}^{\infty} f(x) dx = 1$$

$$\Rightarrow P(a < x < b) = \int_a^b f(x) dx$$

$$\Rightarrow F(x) = P(X \leq x) = \int_{-\infty}^x f(t) dt$$

$$f'(x) = f(x)$$

⇒ Expectation

$$\mu = E(x) = \sum_{\forall x} x p(x)$$

⇒ Variance

$$\begin{aligned} \sigma^2 &= E[(X - \mu)^2] \\ &= E[X^2] - (E[X])^2 \end{aligned}$$

$$\Rightarrow \text{var}(c) = 0$$

$$\text{var}(cn) = c^2 \text{var}(n)$$

\Rightarrow n^{th} moment about origin

$$\mu_r' = E[X^r] = \sum x^r p(x) = \int_{-\infty}^{\infty} x^r f(x) dx$$

\Rightarrow Moment generation function

$$M_X(t) = E[e^{tn}] = \int_{-\infty}^{\infty} e^{tn} f(x) dx$$

\Rightarrow Mean deviation abt mean

$$MD = \sum |x - \mu| p(x) = \int_{-\infty}^{\infty} |x - \mu| f(x) dx$$

\Rightarrow characteristic function

$$\phi_X(t) = E[e^{itn}]$$

\Rightarrow Bernoulli

$$p(x) = p^n q^{1-n}$$

$$E(X) = p$$

$$V(X) = p q$$

$$M_X(t) = q + p e^t$$

\Rightarrow Binomial

$$P(X=x) = {}^n C_x p^x q^{n-x}$$

$$E(X) = np$$

$$V(X) = npq$$

$$M(X) = (q + pe^t)^n$$

$$E(X) = \sum_{\forall x} xp(x)$$

$$= \sum_{x=0}^n x {}^n C_x p^x q^{n-x}$$

$$= \sum x \frac{n!}{x!(n-x)!} p^x q^{n-x}$$

$$= np \sum \frac{x(n-1)!}{x!(n-x)!} p^{x-1} q^{n-x} (q+p)^{n-1}$$

$$= np$$

For variance

$$n(n-1)p^2(q+p)^{n-2} + np$$

$$= np(1-p)$$

$$\Rightarrow \text{skewness} = \frac{1-2p}{\sqrt{npq}}$$

\Rightarrow geometric distribution

$$P(X=n) = p q^{n-1} \\ = q^{n-1} p$$

$$E(X) = \frac{1}{p} \quad \text{var}(X) = \frac{q}{p^2}$$

\Rightarrow Poisson Distribution

$$P(X=x) = \frac{e^{-\lambda} \lambda^x}{x!} \quad x = 0, 1, 2, \dots$$

$$E(X) = V(X) = \lambda$$

$$M_X(t) = e^{\lambda(e^t - 1)}$$

Covariance

$$\text{cov}(X, Y) = E[(X - \mu_x)(Y - \mu_y)]$$

$$E[X] = \mu_x \quad E[Y] = \mu_y$$

$$\boxed{\text{cov}(X, Y) = E[XY] - E[X]E[Y]}$$

$$\Rightarrow \boxed{\text{var}(X + Y) = \text{var}(X) + \text{var}(Y) + 2\text{cov}(X, Y)}$$

Conditional Distribution

$$P(X = x | Y = y) = \frac{P[X = x \text{ \& } Y = y]}{P[Y = y]}$$

$$E[Y|x] = \sum y f_{Y|X}(y)$$

$$\Rightarrow M_x(t) = e^{\mu t + \frac{1}{2} \sigma^2 t^2}$$

$$\Rightarrow \text{Correlation } \rho = \frac{\text{cov}(X, Y)}{\sqrt{\text{var}(X) \text{var}(Y)}}$$

$$r = \frac{n \sum xy - \sum x \sum y}{\sqrt{\left[n \sum x^2 - (\sum x)^2 \right] \left[n \sum y^2 - (\sum y)^2 \right]}}^{1/2}$$

Chi square distribution

$$\sum_{i=1}^n \frac{(X_i - \mu_i)^2}{\sigma^2}$$

$$E[X] = n$$

$$\text{Var}[X] = 2n$$

$$\chi^2 = \sum_{i=1}^n \frac{(O_i - E_i)^2}{E_i}$$

Student t distribution

$$s^2 = \frac{1}{n-1} \sum (x_i - \bar{x})^2$$

$$t = \frac{\bar{x} - \mu_0}{s / \sqrt{n}}$$

F - Distribution

$$F = \frac{x/\sqrt{1}}{y/\sqrt{2}}$$

$$F = \frac{Sx^2}{Sy^2}$$

Hyper geometric dist

$$\frac{{}^n C_x \cdot {}^{n-x} C_{n-x}}{{}^n C_n}$$

⇒ Uniform Distribution

$$f(x) = \frac{1}{b-a} \quad a \leq x \leq b$$

$$\begin{aligned} E[X] &= \frac{b+a}{2} \\ \sigma^2 &= \frac{(b-a)^2}{12} \end{aligned}$$

⇒ Gamma Distribution

$$\mu = \frac{r}{\lambda}$$

$$\sigma^2 = \frac{r}{\lambda^2}$$

⇒ Normal Distribution

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

mean deviation $\frac{\sqrt{2}}{\pi}$

⇒ Standard Normal Variable $\left(\frac{X-\mu}{\sigma} = z \right)$

$$F(-z) = 1 - F(z)$$

Chebyshev

$$P[|X - \mu| < k\sigma] \geq 1 - \frac{1}{k^2}$$