

List of Formulae and Notation

Definitions

Indicator Function: $I_A = 1$ if A occurs, 0 otherwise.

Marginal PMF/PDF: $p_X(x_j) = \sum_k p_{X,Y}(x_j, y_k); \quad f_X(x) = \int_{-\infty}^{\infty} f_{X,Y}(x, y) dy.$

Marginal CDF: $F_X(x) = F_{X,Y}(x, \infty).$

Joint Moments: $\rho_{X,Y} = \frac{\text{cov}(X,Y)}{\sigma_X \sigma_Y}, \quad \text{cov}(X, Y) = E[XY] - E[X]E[Y].$

Discrete Random Variables

Bernoulli: $p_X(1) = p = 1 - p_X(0), \quad E[X] = p, \text{var}[X] = p(1 - p)$

Binomial: $p_X(k) = \binom{n}{k} p^k (1 - p)^{n-k}, \quad k = 0, 1, \dots, n. \quad E[X] = np, \text{var}[X] = np(1 - p)$

Geometric: $p_X(k) = p(1 - p)^{k-1}, \quad k = 1, 2, \dots. \quad E[X] = \frac{1}{p}, \text{var}[X] = \frac{1 - p}{p^2}$

Poisson: $p_X(k) = \frac{\alpha^k}{k!} e^{-\alpha}, \quad k = 0, 1, \dots. \quad E[X] = \alpha = \text{var}[X]$

Continuous Random Variables

Uniform: $f_X(x) = \frac{1}{b - a}, \quad a < x < b. \quad E[X] = \frac{a + b}{2}, \text{var}[X] = \frac{(b - a)^2}{12}$

Exponential: $f_X(x) = \lambda e^{-\lambda x}, \quad x \geq 0. \quad E[X] = \frac{1}{\lambda}, \text{var}[X] = \frac{1}{\lambda^2}$

Gaussian: $f_X(x) = \frac{1}{\sqrt{2\pi}\sigma^2} \exp\left(-\frac{(x - \mu)^2}{2\sigma^2}\right), \quad x \in \mathbb{R}. \quad E[X] = \mu, \text{var}[X] = \sigma^2$

Gaussianity

Q fn.: $Q(x) = \frac{1}{\sqrt{2\pi}} \int_x^{\infty} e^{-t^2/2} dt, \quad Q(x) = 1 - Q(-x)$

CDF: $X \sim \mathcal{N}(\mu, \sigma^2) \Rightarrow P[X > t] = Q\left(\frac{t - \mu}{\sigma}\right)$

Joint PDF: $f_{\mathbf{X}}(\mathbf{x}) = \frac{1}{2\pi\sqrt{\det(\mathbf{C})}} \exp\left[-\frac{1}{2}(\mathbf{x} - \boldsymbol{\mu})^T \mathbf{C}^{-1}(\mathbf{x} - \boldsymbol{\mu})\right], \quad \mathbf{x} \in \mathbb{R}^2$
 where $\mathbf{C} = E[(\mathbf{X} - \boldsymbol{\mu})(\mathbf{X} - \boldsymbol{\mu})^T]$

Result 1: If \mathbf{X} is jointly Gaussian, then $\mathbf{Y} = \mathbf{A}\mathbf{X} + \mathbf{b}$ is jointly Gaussian.

Result 2: If \mathbf{X} is jointly Gaussian, then its components are marginally Gaussian.

Other Useful Results

Bayes: $P(B_j|A) = \frac{P(A|B_j)P(B_j)}{\sum_{k=1}^n P(A|B_k)P(B_k)},$ where $\{B_k\}_{k=1}^n$ is a partition of \mathcal{S} .

Total Prob.: $P(A) = \sum_{k=1}^n P(A|B_k)P(B_k),$ where $\{B_k\}_{k=1}^n$ is a partition of \mathcal{S} .

$E[X] = \sum_{k=1}^n E[X|B_k]P[B_k]; \quad f_X(x) = \sum_{k=1}^n f_X(x|B_k)P[B_k].$

Functions of X : $Y = aX + b \Rightarrow f_Y(y) = \frac{1}{|a|} f_X\left(\frac{y - b}{a}\right)$

$Y = g(X) \Rightarrow f_Y(y) = \sum_{k=1}^n f_X(x_k) \left| \frac{dx_k}{dy} \right|,$ where $g(x_k) = y, \quad k = 1, \dots, n.$

Independence (rv's): X, Y independent $\Leftrightarrow F_{X,Y}(x, y) = F_X(x)F_Y(y),$
 $f_{X,Y}(x, y) = f_X(x)f_Y(y), \quad p_{X,Y}(x, y) = p_X(x)p_Y(y).$

Sum of rv's: X, Y independent $\Rightarrow f_{X+Y}(z) = f_X(z) * f_Y(z).$