

Chap 6

Probability and Distribution

7월 25일

Sample space Ω

event space \mathcal{A} probability P

Probability $\left\{ \begin{array}{l} \text{Discrete} \quad \text{joint} \quad P(X=a_i, Y=y_j) = \frac{n_{ij}}{N} \\ \text{Continuous} \quad P(a \leq X \leq b) = \int_a^b f(x) dx \end{array} \right.$

Probability & F
Cumulative distribution Function

$$F_X(x) = P(X_1 \leq x_1, \dots, X_n \leq x_n)$$

Rule

① Sum rule

$$p(x) = \begin{cases} \sum_{y \in Y} p(x, y) & \text{if } y \text{ is discrete} \\ \int_Y p(x, y) dy & \text{if } y \text{ is continuous} \end{cases}$$

② Product rule

$$p(x, y) = p(y|x)p(x)$$

③ Bayes' theorem

$$p(x|y) = \frac{p(y|x)p(x)}{p(y)}$$

Properties of
PDF

① Mean

$$E_X[g(x)] = \int_X g(x)p(x)dx$$

② Correlation

$$\text{corr}[x, y] = \frac{\text{cov}[x, y]}{\sqrt{V[x]V[y]}} \in [-1, 1]$$

$$E_X[fa(x)] = aE_X[g(x)] + bE_X[h(x)]$$

$$(f(x) = ag(x) + bh(x))$$

③ Covariance

$$\begin{aligned} \text{Cov}_{X,Y}[x, y] &= E_{X,Y}[(x - E_X[x])(y - E_Y[y])] \\ &= E[xy] - E[x]E[y] \end{aligned}$$

④ Variance

$$= E_X[x^2] - (E_X[x])^2$$

$$V_X[x] = \text{Cov}_X[x, x] = E_X[x x^T] - E_X[x]E_X[x]^T$$

Statistical Independence

1) Statistical Independence

$$p(x, y) = p(x)p(y) \Leftrightarrow X, Y \text{ statistical independent}$$

2) Conditional Independence

$$X, Y \text{ conditional independent} \Leftrightarrow p(x, y|z) = p(x|z)p(y|z)$$

$$\Rightarrow \text{uncorrelated } X, Y, \quad \text{Var}[X+Y] = \text{Var}[X] + \text{Var}[Y] \\ \text{Cov}[X, Y] = 0$$

Distribution

1) Gaussian Distribution $X \sim N(\mu, \Sigma) \rightarrow$ Exponential Family.
+ Standard Normal distribution $\mu=0, \Sigma=I$

2) Bernoulli Distribution $\text{Ber}(\mu)$

$$p(x|\mu) = \mu^x(1-\mu)^{1-x} \quad x \in \{0, 1\},$$

$$E[x] = \mu,$$

$$\text{Var}[x] = \mu(1-\mu)$$

3) Binomial Distribution

$$p(m|N, \mu) = \binom{N}{m} \mu^m (1-\mu)^{N-m}$$

$$E[m] = \mu N \quad \text{Var}[m] = N\mu(1-\mu)$$

4) Beta Distribution + Conjugacy