

Intro: Random Experiment: An outcome for each run.

Sample Space Ω : Set of all possible outcomes.

Event: Subsets of Ω .

Prob. of Event A : $P(A) = \frac{\text{Number of outcomes in } A}{\text{Number of outcomes in } \Omega}$

Axioms: $P(A) \geq 0 \forall A \in \Omega$, $P(\Omega) = 1$,

If $A \cap B = \emptyset$, then $P(A \cup B) = P(A) + P(B) \forall A, B \in \Omega$

Cond. Prob. $P(A|B) = \frac{P(A \cap B)}{P(B)}$

* $P(A \cap B) = P(A|B)P(B) = P(B|A)P(A)$

Independence: $P(A|B) = P(A) \Leftrightarrow P(A \cap B) = P(A)P(B)$

Total Prob. Thm: If H_1, H_2, \dots, H_n form a partition of Ω , then $P(A) = \sum_{i=1}^n P(A|H_i)P(H_i)$.

Bayes' Rule: $P(H_k|A) = \frac{P(H_k \cap A)}{P(A)} = \frac{P(A|H_k)P(H_k)}{\sum_{i=1}^n P(A|H_i)P(H_i)}$

*Posteriori: $P(H_k|A)$, Likelihood: $P(A|H_k)$, Prior: $P(H_k)$

1 RV: CDF: $F_X(x) = P[X \leq x]$

PMF: $P_X(x_j) = P[X = x_j] \ j = 1, 2, \dots$

PDF: $f_X(x) = \frac{d}{dx} F_X(x)$

* $P[a \leq X \leq b] = \int_a^b f_X(x) dx$ IS THIS CORRECT?

Cond. PMF: $P_X(x|A) = P[X = x|A] = \frac{P[X=x, A]}{P[A]}$ IS THIS CORRECT?

Cond. PDF: $f_X(x|A) = \frac{f_{X,A}(x,a)}{f_A(a)}$ IS THIS CORRECT?

Exp.: $E[h(X)] = \int_{-\infty}^{\infty} h(x)f_X(x)dx \mid \sum_{k=-\infty}^{\infty} h(k)f_X(k)$

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

Cond. Exp.: $E[X|A] = \int_{-\infty}^{\infty} xf_X(x|A) dx$

2 RVs: Joint PMF: $P_{X,Y}(x, y) = P[X = x, Y = y]$

Joint PDF: $f_{X,Y}(x, y) = \frac{\partial^2}{\partial x \partial y} F_{X,Y}(x, y)$

* $P[(X, Y) \in A] = \int \int_{(x,y) \in A} f_{X,Y}(x, y) dx dy$

Expectation: $E[g(X, Y)] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x, y)f_{X,Y}(x, y) dx dy$

Correlation: $E[XY]$

Covariance: $\text{Cov}[X, Y] = E[(X - \mu_X)(Y - \mu_Y)] = E[XY] - E[X]E[Y]$

Correlation Coeff.: $\rho_{X,Y} = E \left[\left(\frac{X - \mu_X}{\sigma_X} \right) \left(\frac{Y - \mu_Y}{\sigma_Y} \right) \right] = \frac{\text{Cov}[X, Y]}{\sigma_X \sigma_Y}$

Marginal PMF: $P_X(x) = \sum_{j=1}^{\infty} P_{X,Y}(x, y_j)$

Marginal PDF: $f_X(x) = \int_{-\infty}^{\infty} f_{X,Y}(x, y) dy$

Conditional PMF: $P_{X|Y}(x|Y) = P[X = x|Y = y] = \frac{P_{X,Y}(x,y)}{P_Y(y)}$

Conditional PDF: $f_{X|Y}(x|y) = \frac{f_{X,Y}(x,y)}{f_Y(y)}$

Bayes' Rule $f_{Y|X}(y|x) = \frac{f_{X,Y}(x,y)}{f_X(x)} = \frac{f_{X|Y}(x|y)f_Y(y)}{\int_{-\infty}^{\infty} f_{X|Y}(x|y')f_Y(y') dy'}$

* $P_{Y|X}(y|x) = \frac{P_{X,Y}(x,y)}{P_X(x)} = \frac{P_{X|Y}(x|y)P_Y(y)}{\sum_{j=1}^{\infty} P_{X|Y}(x|y_j)P_Y(y_j)}$

Independent: $f_{X|Y}(x|y) = f_X(x) \forall y \Leftrightarrow f_{X,Y}(x, y) = f_X(x)f_Y(y)$

*If independent, then uncorrelated.

Uncorrelated: $\text{Cov}[X, Y] = 0 \Leftrightarrow \rho_{X,Y} = 0$

Orthogonal: $E[XY] = 0$

Conditional Expectation: $E[Y] = E[E[Y|X]]$ or $E[E[h(Y)|X]]$

* $E[E[Y|X]]$ w.r.t. $X \mid E[Y|X]$ w.r.t. Y .

Estimation: