

Probability

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

If A and B are independent $P(A \cap B) = P(A)P(B)$

$$P(A|B) = \frac{P(B|A)P(A)}{P(B)}$$

Expected Value

$$E[x] = \int_{-\infty}^{\infty} xp(x) dx$$

$$E[g(x)] = \int_{-\infty}^{\infty} g(x)p(x) dx$$

Variance

$$Var[x] = E[(x - E[x])^2] = \sigma^2 = \int_{-\infty}^{\infty} (x - E[x])^2 p(x) dx$$

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

Covariance

$$\begin{aligned} cov(X_1, X_2) &= E[(X_1 - m_1)(X_2 - m_2)] \\ &= E[(X_1)(X_2)] - m_1 m_2 \end{aligned}$$

Correlation

$$\rho = \frac{cov(X_1, X_2)}{\sqrt{V(X_1)V(X_2)}}$$

Probability Distribution

Bernoulli

$$f(x) = \begin{cases} q = 1 - p & \text{if } k = 0 \\ p & \text{if } k = 1 \end{cases} \quad E[x] = p \quad Var[X] = p(1 - p)$$

Binomial

$$f(x) = \binom{n}{k} p^k (1 - p)^{n-k} \quad E[x] = np \quad Var[X] = np(1 - p)$$

Poisson

$$f(x) = \frac{\lambda^k e^{-\lambda}}{k!} \quad E[x] = \lambda \quad Var[X] = \lambda$$

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Pareto

$$f(x) = \frac{\alpha x_m^\alpha}{x^{\alpha+1}} \quad E[x] = \quad Var[X] =$$