

Taylor Series

$$\frac{1}{x} = \sum_{n=0}^{\infty} x^n$$

$$e^x = \sum_{n=0}^{\infty} \frac{x^n}{n!}$$

$$\cos x = \sum_{n=0}^{\infty} (-1)^n \frac{x^{2n}}{(2n)!}$$

$$\sin x = \sum_{n=0}^{\infty} (-1)^n \frac{x^{2n+1}}{(2n+1)!}$$

$$\ln(1+x) = \sum_{n=1}^{\infty} (-1)^{n+1} \frac{x^n}{n}$$

Permutations and Combinations

$$P(n, k) = \frac{n!}{(n-k)!}$$

$$C(n, k) = \binom{n}{k} = \frac{n!}{(n-k)!k!}$$

Laplace Transforms

$$F^*(s) = \int_0^{\infty} f(t)e^{-st} dt$$

$$f(t) = \int_0^{\infty} F^*(s)e^{st} ds$$

Convolution Property

$$f(t) * g(t) = \int_0^t f(t-x)g(x)dx \leftrightarrow F^*(s)G^*(s)$$

Z-Transform

Mapping of discrete function f_n into complex function with variable z .

$$F(z) = \sum_{n=0}^{\infty} f_n z^n$$

Probability and Conditional

$$P(A \cup B) = P(A) + P(B) - P(A \cap B)$$

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

A, B are independent if

$$P(A, B) = P(A)P(B)$$

Total Probability

$$P(B) = \sum_i P(A_i)P(B|A_i)$$

Bayes' Rule

$$P(A_i|B) = \frac{P(A_i \cap B)}{P(B)} = \frac{P(A_i)P(B|A_i)}{\sum_j P(A_j)P(B|A_j)} = \frac{P(A_i)P(B|A_i)}{P(B)}$$

PMF (Probability Mass Function)

$$p_X(x) = p(\{s \in \Omega \text{ s.t. } X(s) = x\})$$

$$\sum_x p_X(x) = 1$$

Bernoulli Random Variable

$X = 1$ on success, $X = 0$ on failure.

$$p(X = x) = p, \text{ if } x = 1$$

$$p(X = x) = 1 - p, \text{ if } x = 0$$

Geometric Random Variable

Counts #trials until first success.

$$p_X(x) = (1-p)^{x-1}p, \text{ } x = 1, 2, \dots$$

$$p(X \geq s+1 | X \geq t) = p(X \geq s)$$

Binomial Random Variable

Counts #success in n identical independent experiments.

$$p_X(x) = \binom{n}{x} p^x (1-p)^{n-x}, \text{ when } 0 \leq x \leq n$$

$$p_X(x) = 0, \text{ otherwise}$$

Poisson Random Variable

Model occurrence of event over time interval assuming event happens at rate λ

$$p_X(x) = e^{-\lambda} \frac{\lambda^x}{x!}, \text{ when } x = 0, 1, \dots$$

PDF (Probability Density Function)

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

CDF (Cumulative Distribution Function)

$$F_X(x) = P(X \leq x)$$

$$\lim_{x \rightarrow -\infty} F_X(x) = 0$$

$$\lim_{x \rightarrow \infty} F_X(x) = 1$$

$$P(a < X \leq b) = F_X(b) - F_X(a)$$

Uniform Distribution

$$f_X(x) = \frac{1}{b-a}, \text{ when } a \leq x \leq b$$

$$f_X(x) = 0, \text{ otherwise}$$

Exponential Distribution

Memoryless continuous distribution.

$$f_X(x) = \lambda e^{-\lambda x}, \text{ when } x \geq 0$$

$$F_X(x) = 1 - e^{-\lambda x}, \text{ when } x \geq 0$$

$$F_X(x) = 0, \text{ otherwise}$$

$$P(X > x) = e^{-\lambda x}$$

Expectation

$$E[X] = \sum_x x p(x)$$

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

$$\text{If } Y = g(X), E[Y] = \sum_x g(x) p(x),$$

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

$$E[X + Y] = E[X] + E[Y]$$

$$E[aX] = aE[X]$$

$$E[XY] = E[X]E[Y], \text{ if } X, Y \text{ are independent.}$$

For X, Y with joint PMF $p(x, y)$ or PDF

$$f_{X,Y}(x, y), E[XY] = \sum_{(x,y)} xy p(x, y),$$

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

Conditional Expectation

$$X, Y \text{ are random variables, } E[Y|X] = \sum_y y P(Y = y | X = x) = \sum_y y p_{Y|X}(y|x),$$

$$E[Y|X] = \int_{-\infty}^{\infty} y f_{Y|X}(y|x) dy$$

Unconditional Expectation

$$E[Y] = \sum_x E[Y|X] p_X(x)$$

$$E[Y] = \int_{-\infty}^{\infty} E[Y|X] f_X(x) dx$$

Variance

$$\text{Var}[X] = E[(X - E[X])^2] = \sum_x (x - E[X])^2 p(x) = \int_{-\infty}^{\infty} (x - E[X])^2 f_X(x) dx$$

$$\text{Var}[X] = E[X^2] - E[X]^2$$

$$\text{Var}[X + Y] = \text{Var}[X] + \text{Var}[Y], \text{ if } X, Y \text{ are independent.}$$

Expectations and Variances

$$\text{Binomial: } np, np(1-p)$$

$$\text{Geometric: } \frac{1}{p}, \frac{1-p}{p^2}$$

$$\text{Uniform: } \frac{a+b}{2}, \frac{(b-a)^2}{12}$$

$$\text{Exponential: } \frac{1}{\lambda}, \frac{1}{\lambda^2}$$

$$\text{Poisson: } \lambda, \lambda$$

Covariance: measure of joint probability

$$\text{Cov}(X, Y) = E[(X - E[X])(Y - E[Y])]$$

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

If X, Y are independent, $\text{Cov}(X, Y) = 0$

Correlation: scaled version of covariance

$$\rho(X, Y) = \frac{\text{Cov}(X, Y)}{\sqrt{\text{Var}(X)\text{Var}(Y)}}, \text{ range } [-1, 1]$$

Q1: A given program has an execution time that is uniformly distributed between 10 and 20 seconds. The number of interrupts that occur during execution is a Poisson random variable with parameter t where t is the program execution time. The probability distribution of the number of interrupts is therefore $P(N = k) = (\lambda t) k e^{-\lambda t}$.

(a) What is $E[N|T = t]$, where N is the number of interrupts the program experiences, and T is the running time of the program.

$E[N|T = t] = \lambda t$, since for fixed running time, the number of interrupts is a Poisson random variable with mean λt .

(b) Find the expected number of interrupts the program experiences during a randomly selected execution.

$$E[N] = \int_{10}^{20} E[N|T = t] f_T(t) dt = \int_{10}^{20} \lambda t \frac{1}{10} dt = 15\lambda$$

Q2: Suppose that you made a webpage and you are collecting the statistics from the visitors. There are m types of visitors. Each visit is equally likely to be any of the m types. Find the expected number of visitors needed in order to have at least one of each type. Hint: Let X denote the number of visitors needed. It is useful to represent X by $X = \sum_{i=1}^m X_i$ where each X_i is a geometric random variable.

Suppose the current visitor pool contains i different types. Let X_i denote the number of additional visitors needed until it contains $i+1$ types. The X_i are independent geometric random variables with parameter $(m-i)/m$, $i = 0, 1, \dots, m-1$. $E[X] = E[\sum_{i=1}^m X_i] = \sum_{i=1}^m E[X_i] = \sum_{i=1}^m \frac{m}{m-i}$

Q3: A Markov chain $\{X_n, n \geq 0\}$ with states 0, 1, 2, has the transition

probability matrix $\begin{bmatrix} \frac{1}{2} & \frac{1}{3} & \frac{1}{6} \\ 0 & \frac{1}{3} & \frac{2}{3} \\ \frac{1}{2} & 0 & \frac{1}{2} \end{bmatrix}$ If

$P[X_0 = 0] = P[X_0 = 1] = 1$, find the state probability vector $P[X_3 = 2]$.

Cubing the transition probability matrix,

we obtain $P^3 = \begin{bmatrix} \frac{13}{36} & \frac{11}{54} & \frac{47}{108} \\ \frac{4}{9} & \frac{27}{2} & \frac{11}{27} \\ \frac{5}{12} & \frac{9}{36} & \frac{13}{36} \end{bmatrix}$

$$P[X_3 = 2] = \frac{1}{4} \cdot \frac{47}{108} + \frac{1}{4} \cdot \frac{11}{27} + \frac{1}{2} \cdot \frac{13}{36}$$

Q4: A workstation tries to transmit frames through Ethernet. Suppose that whether or not collision occurs in the current transmission depends on the result of the last two transmissions the workstation had. That is, suppose that if collisions have occurred in both of the past two transmissions, then with probability 0.7 a collision will occur in the current transmission; if a collision occurs in last transmission but not the transmission before the last one, then a collision will occur in the current transmission with probability 0.5; if a collision occurred in the transmission before the last one but not the last one, then one will occur in the current transmission with probability 0.4; if there have been no collision in the past two transmissions, then a collision will occur in the current transmission with probability 0.2. (Hint: Note that the state description needs to include status of last two transmissions).

(b) Find the transition probability matrix.

$$P = \begin{bmatrix} 0.7 & 0 & 0.3 & 0 \\ 0.5 & 0 & 0.5 & 0 \\ 0 & 0.4 & 0 & 0.6 \\ 0 & 0.2 & 0 & 0.8 \end{bmatrix}$$

(c) What fraction of frames suffer a collision?

Solve $\pi = \pi P$ to obtain the stationary state probabilities. Then the fraction of frames suffering a collision is $\pi_0 + \pi_2$ (or $\pi_0 + \pi_1$).