## 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 fuction 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

$$\begin{array}{l} P(A_i|B) = \frac{P(A_i \cap B)}{P(B)} = \frac{P(A_i)P(B|A_i)}{P(B)} = \\ \frac{P(A_i)P(B|A_i)}{\sum_i P(A_j)P(B|A_j)} \end{array}$$

#### 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.

$$p(X = x) = p$$
, if  $x = 1$ 

$$p(X = x) = 1 - p$$
, if  $x = 0$ 

# Geometric Random Variable

Counts #trials until first success.

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

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

#### Binomial Random Variable

Counts #success in n identical independent experiments

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

 $p_X(x) = 0$ , otherwise

# Poisson Random Variable

Model occurrence of event over time interval assuming event happens at rate  $\lambda$ 

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

# PDF (Probability Density Function)

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

# CDF (Cumulative Distribution Function)

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

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

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

$$P(a < X \le b) = F_X(b) - F_X(a)$$
  
Uniform Distribution

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

 $f_X(x) = 0$ , otherwise

#### **Exponential Distribution**

Memoryless continuous distribution.

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

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

$$F_X(x) = 0$$
, otherwise

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

# Expectation

$$E[X] = \sum_{x} x p(x)$$

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

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], if X,Y are independent.

For X, Y with joint PMF p(x, y) or PDF  $f_{X,Y}(x, y), E[XY] = \sum_{(x,y)} xyp(x, y),$   $E[XY] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} xyf_{X,Y}(x, y)dxdy$ 

#### Conditional Expectation

X,Y are random variables,

$$\begin{split} E[Y|X] &= \sum_{y} y P(Y=y|X=x) = \sum_{y} y p_{Y|X}(y|x), \\ E[Y|X] &= \int_{-\infty}^{\infty} y f_{Y|X}(y|x) dy \end{split}$$

#### Unconditional Expectation

$$E[Y] = \sum_{x} E[Y|X]p_X(x)$$

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

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

$$Var[X] = E[X^2] - E[X]^2$$

Var[X + Y] = Var[X] + Var[Y], if X,Y are independent.

#### **Expectations and Variances**

Binomial: np, np(1-p)

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

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

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

Poisson:  $\lambda$ ,  $\lambda$ 

Covariance: measure of joint probability

$$Cov(X,Y) = E[(X - E[X])(Y - E[Y])]$$

$$Cov(X,Y) = E[XY] - E[X]E[Y]$$

If X,Y are independent, Cov(X,Y) = 0

Correlation: scaled version of covariance

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

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  $P(N = k) = (\lambda t)ke^{\lambda t}$ .

(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  $\lambda t$ .

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

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

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  $X_i$ is 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

$$\begin{bmatrix} \frac{1}{2} & \frac{1}{3} & \frac{1}{6} \\ 0 & \frac{1}{3} & \frac{2}{3} \\ \frac{1}{2} & 0 & \frac{1}{2} \end{bmatrix} \text{ If } P[X_0 = 0] = P[X_0 = 1] = 1, \text{ find}$$

the state probability vector  $P[X_3 = 2]$ .

Cubing the transition probability matrix, we obtain

$$P^{3} = \begin{bmatrix} \frac{13}{36} & \frac{11}{54} & \frac{47}{108} \\ \frac{4}{9} & \frac{47}{27} & \frac{11}{27} \\ \frac{5}{12} & \frac{2}{9} & \frac{13}{36} \end{bmatrix}$$
$$P[X_{3} = 2] = \frac{1}{4} \cdot \frac{47}{108} + \frac{1}{4} \cdot \frac{11}{27} + \frac{1}{2} \cdot \frac{13}{36}$$

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 trans- missions 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  $\pi_0 + \pi_1$ ).