

Intro. to Stochastic Processes

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Outline

- Stochastic Process
- Counting Process
- Poisson Process
- Markov Process

P. 2

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Stochastic Process

- A stochastic process $\underline{N} = \{N(t), t \in T\}$ is a collection of r.v., i.e., for each t in the index set T , $N(t)$ is a random variable
 - t : time
 - $N(t)$: state at time t
 - If T is a countable set, \underline{N} is a discrete-time stochastic process
 - If T is continuous, \underline{N} is a continuous-time stoc. proc.

P. 3

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Counting Process

- A stochastic process $\{N(t), t \geq 0\}$ is said to be a counting process if $N(t)$ is the total number of events that occurred up to time t . Hence, some properties of a counting process is
 - $N(t) \geq 0$
 - $N(t)$ is integer valued
 - If $s < t$, $N(t) \geq N(s)$
 - For $s < t$, $N(t) - N(s)$ equals number of events occurring in the interval $(s, t]$

P. 4

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Counting Process

- Independent increments
 - If the number of events that occur in disjoint time intervals are independent
- Stationary increments
 - If the dist. of number of events that occur in any interval of time depends only on the length of time interval



P. 5

Poisson Process

- Def. A: the counting process $\{N(t), t \geq 0\}$ is said to be Poisson process having rate $\lambda, \lambda > 0$ if
 - $N(0) = 0$;
 - The process has independent-increments
 - Number of events in any interval of length t is Poisson dist. with mean λt , that is for all $s, t \geq 0$.

$$P[N(t+s) - N(s) = n] = e^{-\lambda t} \frac{(\lambda t)^n}{n!}$$

$$n = 0, 1, 2, \dots$$



P. 6

Poisson Process

- Def. B: The counting process $\{N(t), t \geq 0\}$ is said to be a Poisson process with rate $\lambda, \lambda > 0$, if
 - $N(0) = 0$
 - The process has stationary and independent increments
 - $P[N(h) = 1] = \lambda h + o(h)$
 - $P[N(h) \geq 2] = o(h)$
 - The func. f is said to be $o(h)$ if $\lim_{h \rightarrow 0} \frac{f(h)}{h} = 0$
 - Def A \Leftrightarrow Def B, i.e., they are equivalent.
 - We show Def B \Rightarrow Def A
 - Def A \Rightarrow Def B is HW



P. 7

Important Properties

- Property 1: mean number of event for any $t \geq 0$, $E[N(t)] = \lambda t$.
- Property 2: the inter-arrival time dist. of a Poisson process with rate λ is an exponential dist. with parameter λ .
- Property 3: the superposition of two independent Poisson process with rate λ_1 and λ_2 is a Poisson process with rate $\lambda_1 + \lambda_2$



P. 8

Properties (cont.)

- Property 4: if we perform Bernoulli trials to make independent random erasures from a Poisson process, the remaining arrivals also form a Poisson process
- Property 5: the time until r th arrival, i.e., τ_r is known as the r th order waiting time, is the sum of r independent exponential values of τ and is described by Erlang pdf.



P. 9

Markov Process

- $P[X(t_{n+1}) \leq X_{n+1} | X(t_n) = x_n, X(t_{n-1}) = x_{n-1}, \dots, X(t_1) = x_1] = P[X(t_{n+1}) \leq X_{n+1} | X(t_n) = x_n]$
 - Probabilistic future of the process depends only on the current state, not on the history
 - We are mostly concerned with discrete-space Markov process, commonly referred to as Markov chains
 - Discrete-time Markov chains
 - Continuous-time Markov chains



P. 10

DTMC

- Discrete Time Markov Chain:
 - $P[X_{n+1} = j | X_n = k_n, X_{n-1} = k_{n-1}, \dots, X_0 = k_0] = P[X_{n+1} = j | X_n = k_n]$
- discrete time, discrete space
- A finite-state DTMC if its state space is finite
- A homogeneous DTMC if $P[X_{n+1} = j | X_n = i]$ does not depend on n for all i, j , i.e., $P_{ij} = P[X_{n+1} = j | X_n = i]$, where P_{ij} is one step transition prob.



P. 11

Definition

- $P = [P_{ij}]$ is the transition matrix

$$P = \begin{bmatrix} p_{00} & p_{01} & \dots & p_{0j} & \dots \\ p_{10} & p_{11} & \dots & p_{1j} & \dots \\ \dots & \dots & \dots & \dots & \dots \\ p_{i0} & \dots & \dots & p_{ij} & \dots \\ \dots & \dots & \dots & \dots & \dots \end{bmatrix}$$

where $p_{ij} \geq 0$ and $\sum_j p_{ij} = 1$

- A matrix that satisfies those conditions is called a stochastic matrix
- n -step transition prob.
 - $p_{ij}^n = P[x_n = j | x_0 = i]$
 - $i, j, n \geq 0$, p_{ij}^n is the prob. of going from state i to j in n step



P. 12

Chapman-Kolmogorov Eq.

□ Def.

For all $n \geq 0, m \geq 0, i, j \in I$

$$p_{ij}^{(n+m)} = \sum_{k \in I} p_{ik}^n p_{kj}^m$$

in matrix form $P^{n+m} = P^n P^m$ where $P^n = [p_{ij}^n]$

□ Proof:



P. 13

Question

- We have only been dealing with conditional prob. but what we want is to compute the unconditional prob. that the system is in state j at time n , i.e.

$$\pi_n(j) = p(x_n = j)$$

So, given the initial dist. of x_0 , i.e.,

$$\pi_0(i) = p(x_0 = i) \text{ and } \sum_{i \in I} \pi_0 = 1$$

we can get

$$\begin{aligned} p[x_n = j] &= \sum_{i \in I} p(x_n = j | x_0 = i) \pi_0(i) \\ &= \sum_{i \in I} p_{ij}^n \pi_0(i) \end{aligned}$$



P. 14

Result 1

- For all $n \geq 1, \pi_n = \pi_0 P^n$, where $\pi_m = (\pi_m(0), \pi_m(1), \dots)$ for all $m \geq 0$. From the above equ., we deduce that $\pi_{n+1} = \pi_n P$. Assume that $\lim_{n \rightarrow \infty} \pi_n(i)$ exists for all i , and refer it as $\pi(i)$. The remaining question is how to compute π

– Reachable: a state j is reachable from i . If

$$p_{ij}^n > 0 \text{ for some } n \geq 1$$

– Communicate: if j is reachable from i and if i is reachable from j , then we say that i and j communicate ($i \leftrightarrow j$)



P. 15

Result 1 (cont.)

- Irreducible:

– A M.C. is irreducible if $i \leftrightarrow j$ for all $i, j \in I$

- Aperiodic:

– For every state $i \in I$, define $d(i)$ to be largest common divisor of all integer n , s.t.,

$$p_{ij}^n > 0 \text{ if } d(i) = 1 \text{ then the state is aperiodic}$$



P. 16

Result 2

- Invariant measure of a M.C., if a M.C. with transition matrix P is irreducible and aperiodic and if the system of equation $\pi = \pi P$ and $\pi \mathbf{1} = 1$ has a strict positive solution then $\pi(i) = \lim_{n \rightarrow \infty} \pi_n(i)$ independently of initial dist.
 - Invariant equ. : $\pi = \pi P$
 - Invariant measure π



P. 17

CTMC

- Continuous-time Markov Chain
 - Continuous time, discrete state
 - $P[X(t) = j \mid X(s) = i, X(s_{n-1}) = i_{n-1}, \dots, X(s_0) = i_0] = P[X(t) = j \mid X(s) = i]$
 - A continuous M.C. is homogeneous if
 - o $P[X(t+u) = j \mid X(s+u) = i] = P[X(t) = j \mid X(s) = i] = P_{ij}[t-s]$, where $t > s$
 - Chapman-Kolmogorov equ.

For all $t > 0, s > 0, i, j \in I$

$$P_{ij}(t+s) = \sum_{k \in I} P_{ik}(t) P_{kj}(s)$$



P. 18

CTMC (cont.)

- $\pi(t) = \pi(0)e^{Qt}$
 - Q is called the infinitesimal generator
 - Proof:



P. 19

Result 3

- If a continuous M.C. with infinitesimal generator Q is irreducible and if the system of equations $\pi Q = 0$, and $\pi \mathbf{1} = 1$, has a strictly positive solution then $\pi(i) = \lim_{t \rightarrow \infty} p(x(t) = i)$ for all $i \in I$, independently of the initial dist.



P. 20