

01:XXX:XXX - Homework n

Pranav Tikkawar

October 25, 2024

Missed notes: Counting processes
 $\{N(t), t \geq 0\}$ They follow 3 properties:

1. $N(t) \geq 0$
2. $N(t)$ is integer valued
3. $N(t)$ is monotone increasing

$$N(t) : \mathbb{R} \rightarrow \mathbb{N}$$

Monotone increasing function of t

$$N(t) - N(s) = \text{Number of events in } (t, s]$$

Little o notation

A function f is said to be little o ($o(h)$) if

$$\lim_{h \rightarrow 0} \frac{f(h)}{h} = 0$$

eg: $f(h) = h^2$ is little $o(h)$

If u add two function in little $o(h)$ then it is still little $o(h)$

Definition: A counting process $\{N(t), t \geq 0\}$ is a Poisson process if:

1. $N(0) = 0$
2. The number of events in disjoint intervals are independent.
3. $P(N(t+h) - N(t) = 1) = \lambda h + o(h)$ where λ is the rate of the Poisson process. (this mean it is dependant on the length of the interval)
4. $P(N(t+h) - N(t) \geq 2) = o(h)$

Lemma 5.1:

Let $\{N(t), t \geq 0\}$ be a Poisson process. Define $\{N_s(t), t \geq 0\}$ by $N_s(t) = N(s+t) - N(s)$
 Then $\{N_s(t), t \geq 0\}$ is a Poisson process with rate λ

Proof:

$$N_s(0) = N(s+0) - N(s) = 0$$

$$(a, b) \cap (c, d) = \emptyset$$

$$P(N_s(b) - N_s(a) = x, N_s(d) - N_s(c) = y)$$

$$P(N(b-s) - N(a-s) = x, N(d-s) - N(c-s) = y)$$

$$P(N(b-s) - N(a-s) = x)P(N(d-s) - N(c-s) = y)$$

$$P(N_s(b) - N_s(a) = x)P(N_s(d) - N_s(c) = y)$$

Thus disjoint intervals are independent.

$$P(N_s(t+h) - N_s(t) = 1) = P(N(s+t+h) - N(s+t) = 1)$$

We assume N has stationary increments.

$$P(N(s+t+h) - N(s+t) = 1) = P(N(t+h) - N(t) = 1) = \lambda h + o(h)$$

Lmma 5.2:

Let $T_1 = \min(t > 0 : N(t) = 1)$

it is time of arrival

T_1 is exponentially distributed with rate λ

Proof:

$$P_0(t) = P(N(t) = 0)$$

$$P_0(t+h) = P(N(t) = 0, N(t+h) - N(t) = 0)$$

$$P_0(t+h) = P(N(t) = 0)P(N(t+h) - N(t) = 0)$$

$$P_0(t+h) = P_0(t)(1 - \lambda h - o(h))$$

note that $-2o(h) = o(h)$ cuz it basically 0

$$P_0(t+h) = P_0(t) - \lambda h P_0(t) + o(h)P_0(t)$$

$$\frac{dP_0(t)}{dt} = -\lambda P_0(t) + 0$$

This solves to with IC $P_0(0) = 1$

$$P_0(t) = e^{-\lambda t}$$

Define:

T_n for $n \geq 1$ is the time between the $(n-1)th$ and nth arrival.

Proposition 5.4:

T_1, T_2, \dots are independent and exponentially distributed with rate λ

Proof:

Rea book.

Remark:

Define $S_n = \sum_{i=1}^n T_i$

From last time, S_n has a gamma distribution with parameters n and λ

$$f_{S_n}(t) = \frac{\lambda^n t^{n-1} e^{-\lambda t}}{(n-1)!}$$

Theorem 5.1

If $\{N(t), t \geq 0\}$ is a Poisson process with parameter λ then $N(t)$ is a poisson random variable with parameter λt

Proof:

$$\begin{aligned}
P(N(t) = n) &= \int_0^\infty P(N(t) = n | S_n = t) \frac{\lambda^n t^{n-1} e^{-\lambda t}}{(n-1)!} dt \\
&= P(T_{n+1} = t - s | T_1 + T_2 + \dots + T_n = s) \\
&= P(T_{n+1} = t - s) \\
&= \frac{(\lambda t)^n e^{-\lambda t}}{n!}
\end{aligned}$$

Example

Let $\{N(t), t \geq 0\}$ be a Poisson process with rate $\lambda = \frac{1}{3}$

Find:

a) $P(N(5) > N(3))$

This means there are > 0 events in $(3, 5]$

$$\begin{aligned}
P(N(5) > N(3)) &= 1 - P(N(5) - N(3) = 0) \\
&= 1 - P(N(2) = 0) \\
&= 1 - e^{-\frac{2}{3}}
\end{aligned}$$

b) $P(\{N(4) = 1\}, \{N(5) = 3\})$

c) $E(N(5) | N(3) = 2)$

d) $E(T_b | N(3) = 4)$

Last time we finished 5.3.2 + examples

5.3.3 Further thinning of a poisson process.

Suppose $\{N(t), t \geq 0\}$ is a Poisson process with rate λ

There are events of 2 types: 1 w/ probability p and 2 w/ probability $1 - p$

Write $N_1(t)$ for the number of type 1 events in $(0, t]$

$N_2(t)$ for the number of type 2 events in $(0, t]$

Proposition 5.5

$\{N_1(t), t \geq 0\}$ is a Poisson process with rate $p\lambda$ and $\{N_2(t), t \geq 0\}$ Poisson process with rate $(1 - p)\lambda$

Compound Poisson process

Suppose random variables are iid with distribution F with mean μ and variance σ^2

The non-negative integer valued random variable $S = \sum_{i=1}^N X_i$ is called a compound Poisson random variable.

Conditional Variance formula

$$Var(Y) = E(Var(Y|X)) + Var(E(Y|X))$$

If N is a poisson random variable with parameter λ then:

$$Var(S) = \lambda\sigma^2 + \mu^2\lambda$$

Read example 5.27

1 add Missed info

yes

2 10/25

$$P_j = \lim_{t \rightarrow \infty} P_{ij}(t)$$

$$0 = \sum_{k \neq j} q_{kj} P_k - v_j P_j$$

Read remarks of page 395

Example. Limiting probability: P_j for the Birth-Death process with birth rate λ_j and death rate μ_j

Write the balance equations for P_j

$$\lambda_0 P_0 = \mu_1 P_1$$

$$(\lambda_1 + \mu_1) P_1 = \lambda_0 P_0 + \mu_2 P_2$$

$$(\lambda_n + \mu_n) P_n = \lambda_{n-1} P_{n-1} + \mu_{n+1} P_{n+1}$$

We can now go into canceling.

$$\lambda_1 P_1 = \mu_2 P_2$$

$$P_2 = \frac{\lambda_1}{\mu_2} P_1 = \frac{\lambda_1 \cdot \lambda_0}{\mu_2 \cdot \mu_1}$$

$$P_n = \frac{\lambda_{n-1} \cdot \lambda_{n-2} \cdots \lambda_0}{\mu_{n+1} \cdot \mu_n \cdots \mu_1} P_0$$

Use $\sum_{j=0}^{\infty} P_j = 1$ to find limiting probability

$$1 = P_0 + P_0 \sum_{n=1}^{\infty} \frac{\lambda_{n-1} \cdot \lambda_{n-2} \cdots \lambda_0}{\mu_{n+1} \cdot \mu_n \cdots \mu_1}$$

$$P_0 = \frac{1}{1 + \sum_{n=1}^{\infty} \frac{\lambda_{n-1} \cdot \lambda_{n-2} \cdots \lambda_0}{\mu_{n+1} \cdot \mu_n \cdots \mu_1}}$$

We need the infinite sum to be finite for it to be a valid probability.

Read Examples 6.13, 6.14, 6.15 and skip 6.16

Chapter 6.6 Time reversibility

If limiting probabilities exist, a CTMC is called ergodic.

Consider an ergodic CTMC that has been running a long time.

First look at the embedded discrete time markov chain.(forget the time spent in each state.

Just look at the transitions)

Let π_i be the limiting probability of being in state i of the embedded chain.

Recall that $\pi_1 = \sum_j \pi_j P_{ji}$ and $\sum_i \pi_i = 1$

Note that π_i is the proportion of transitions

Recall that $\frac{1}{v_i}$ mean time spent in state i

Claim P_i is the proportion of time the CTMC is in state i more precisely $P_i = \frac{\pi_i/v_i}{\sum_j \pi_j/v_j}$

Verification:

Know $v_i P_i = \sum_{j \neq i} P_j q_{ji} = \sum_{j \neq i} P_j v_j P_{ji} = \sum_j P_j v_j P_{ji}$ Through some more manipulation we can see that this can only be satisfied by π_i

Time reversibility:

Reversing a CTMC Assume the process has been running a long time. Observe it backwards.

Pg 402