STAT253/317 Lecture 15

Yibi Huang

7.3. Limit Theorems

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Let $\{N(t), t \ge 0\}$ be a renewal process with i.i.d interarrival times X_i , i = 1, 2, ... and $\mathbb{E}[X_i] = \mu$.

Explicit forms of N(t) and $m(t) = \mathbb{E}[N(t)]$ are usually *unavailable*. However the limiting behavior of N(t) and m(t) is useful and intuitively makes sense.

As
$$t \to \infty$$
,
 $\blacktriangleright \frac{N(t)}{t} \to \frac{1}{\mu}$ with probability 1 (Proposition 7.1)
 $\flat \frac{m(t)}{t} \to \frac{1}{\mu}$ (Thm 7.1 Elementary Renewal Theorem)

<u>Remark.</u>

▶ The number $1/\mu$ is called the **rate** of the renewal process

▶ Theorem 7.1 is not a simple consequence of Proposition. 7.1, since $X_n \to X$ w/ prob. 1 does not ensure $\mathbb{E}[X_n] \to \mathbb{E}[X]$.

 $X_n \to X$ Does Not Ensure $\mathbb{E}[X_n] \to \mathbb{E}[X]$

Example 7.8 Let U be a random variable which is uniformly distributed on (0, 1); and define the random variables X_n , $n \ge 1$, by

$$X_n = \begin{cases} 0 & \text{if } U > 1/n \\ n & \text{if } U \le 1/n \end{cases}$$

Then $P(X_n = 0) = P(U > 1/n) = 1 - 1/n \rightarrow 0$ as $n \rightarrow \infty$. So with probability 1

$$X_n \to X = 0.$$

However,

$$\mathbb{E}[X_n] = 0 \mathbb{P}(X_n = 0) + n \mathbb{P}(X_n = n) = n imes rac{1}{n} = 1$$
 for all $n \ge 1$.

and hence $\lim_{n\to\infty} \mathbb{E}[X_n] = 1 \neq \mathbb{E}[X] = \mathbb{E}[0] = 0.$

Proof of Proposition 7.1

$$N(t) = n$$

Since
$$S_{N(t)} \leq t < S_{N(t)+1}$$
, we know

$$\frac{S_{N(t)}}{N(t)} \leq \frac{t}{N(t)} < \frac{S_{N(t)+1}}{N(t)}$$
.
By SLLN, $\frac{S_{N(t)}}{N(t)} = \frac{\sum_{i=1}^{N(t)} X_i}{N(t)} \rightarrow \mu$ as $N(t) \rightarrow \infty$, we obtain
 $\frac{S_{N(t)}}{N(t)} \rightarrow \mu$ as $t \rightarrow \infty$. Furthermore, writing
 $\frac{S_{N(t)+1}}{N(t)} = \frac{S_{N(t)+1}}{N(t)+1} \times \frac{N(t)+1}{N(t)}$

we have that ${\cal S}_{{\it N}(t)+1}/({\it N}(t)+1) \rightarrow \mu$ by the same reasoning as before and

$$rac{N(t)+1}{N(t)}
ightarrow 1 ext{ as } t
ightarrow \infty \quad ext{since } P(\lim_{t
ightarrow \infty} N(t) = \infty) = 1$$

Hence, $S_{N(t)+1}/N(t)
ightarrow \mu$.

Stopping Time

Definition. Let $\{X_n : n \ge 1\}$ be a sequence of independent random variables. An integer-valued random variable N > 0 is said to be a *stopping time* w/ respect to $\{X_n : n \ge 1\}$ if the event $\{N = n\}$ is independent of $\{X_k : k \ge n+1\}$.

Example. (*Independent case.*) {N=n} only depends on X1, X2,..., Xn If N is independent of $\{X_n : n \ge 1\}$, then N is a stopping time.

Example. (*Hitting Time* I.) For any set A, the first time X_n hits set A, $N_A = \min\{n : X_n \in A\}$, is a stopping time because

$$\{N_A = n\} = \{X_i \notin A \text{ for } i = 1, 2, \dots, n-1, \text{ but } X_n \in A\}$$

is independent of $\{X_k : k \ge n+1\}$.

Example. (*Hitting Time* II.) For $n \ge 1$, let $S_n = \sum_{k=1}^n X_k$. For any set A, $N_A = \min\{n : S_n \in A\}$, the first time S_n hits set A, is also a stopping time w/ respect to $\{X_n : n \ge 1\}$ because $\{N_A = n\} = \{\sum_{k=1}^i X_k \notin A \text{ for } 1 \le i \le n-1, \text{ but } \sum_{k=1}^n X_k \in A\}$ is independent of $\{X_k : k \ge n+1\}$. Lecture 15 - 5

Example of Non-Stopping Times

(Last visit time) The last time that X_n visit a set A

 $N_A = \max\{n : X_n \in A\}$

is NOT a stopping time.

Clearly we need to know whether A will be visited again in the future to determine such a time. your net earning

▶ The time X_n reaches its maximum,

$$N = \min\{n : X_n = \max_{k \ge 1} X_k\},$$

is NOT a stopping time since

$$\{N = n\} = \{X_n > X_k \text{ for } 1 \leq k < n \text{ and } k \geq n+1\}$$

depends on $\{X_k : k \ge n+1\}$.

Renewal Processes and Stopping Times

Consider a renewal process N(t). With respect to its interarrival times X_1, X_2, \ldots ,

• N(t) is NOT a stopping time.

$$N(t) = n \Leftrightarrow X_1 + \dots + X_n \leq t \text{ and } X_1 + \dots + X_{n+1} > t,$$

depends on X_{n+1} .

• But N(t) + 1 is a stopping time, since

$$N(t) + 1 = n \Leftrightarrow N(t) = n - 1$$

 $\Leftrightarrow X_1 + \dots + X_{n-1} \le t \text{ and } X_1 + \dots + X_n > t,$

is independent of X_{n+1}, X_{n+2}, \ldots

Wald's Equation For a fair game, $E[X_i] = 0$,

If $X_1, X_2...$ are i.i.d. with $\mathbb{E}[X_i] < \infty$, and if N is a stopping time for this sequence with $\mathbb{E}[N] < \infty$, then

$$\mathbb{E}\left[\sum_{j=1}^{N} X_j\right] = \mathbb{E}[N]\mathbb{E}[X_1]$$

Proof. Let us define the indicator variable

$$I_j = egin{cases} 1 & ext{if } j \leq N \ 0 & ext{if } j > N. \end{cases}$$

E[X1 |N=n] may not be equal to E[X1].

We have

$$\sum_{j=1}^{N} X_j = \sum_{j=1}^{\infty} X_j I_j$$

Hence

justified if
$$\mathbb{E}[|X_i|]$$
 is finite

$$\mathbb{E}\left[\sum_{j=1}^{N} X_j\right] = \mathbb{E}\left[\sum_{j=1}^{\infty} X_j I_j\right] = \sum_{j=1}^{\infty} \mathbb{E}[X_j I_j] \qquad (1)$$

Proof of Wald's Equation (Cont'd)

Note I_j and X_j are independent because

$$I_j = 0 \quad \Leftrightarrow \quad N < j \quad \Leftrightarrow \quad N \le j - 1$$

and the event $\{N \leq j-1\}$ depends on X_1, \ldots, X_{j-1} only, but not X_j . From (1), we have

$$\mathbb{E}\left[\sum_{j=1}^{N} X_{j}\right] = \sum_{j=1}^{\infty} \mathbb{E}[X_{j}I_{j}] = \sum_{j=1}^{\infty} \mathbb{E}[X_{j}]\mathbb{E}[I_{j}]$$
$$= \mathbb{E}[X_{1}]\sum_{j=1}^{\infty} \mathbb{E}[I_{j}] = \mathbb{E}[X_{1}]\sum_{j=1}^{\infty} P(N \ge j)$$
$$= \mathbb{E}[X_{1}]\mathbb{E}[N]$$

Here we use the alternative formula $\mathbb{E}[N] = \sum_{j=1}^{\infty} P(N \ge j)$ to find expected values of non-negative integer valued random variables.

Proposition 7.2

$$\mathbb{E}[S_{N(t)+1}] = \mu(m(t)+1)$$

Proof: Since N(t) + 1 is a stopping time, by Wald's equation, we have

$$\mathbb{E}[S_{N(t)+1}] = \mathbb{E}\left[\sum_{j=1}^{N(t)+1} X_j\right] = \mathbb{E}[N(t)+1]\mathbb{E}[X_1] = (m(t)+1)\mu$$

Since $S_{N(t)+1} = t + Y(t)$, where Y(t) is the residual life at t, taking expectations and using the result above yields

$$\mathbb{E}[S_{N(t)+1}] = \mu(m(t)+1) = t + \mathbb{E}[Y(t)].$$

So far we have proved Proposition 7.2 and can deduce that

$$rac{m(t)}{t} = rac{1}{\mu} - rac{1}{t} + rac{\mathbb{E}[Y(t)]}{t\mu}$$

Proof of the Elementary Renewal Theorem

First from Proposition 7.2, we have
$$\frac{m(t)}{t} = \frac{1}{\mu} - \frac{1}{t} + \frac{\mathbb{E}[Y(t)]}{t\mu} \ge \frac{1}{\mu} - \frac{1}{t} \quad \Rightarrow \quad \lim_{t \to \infty} \frac{m(t)}{t} \ge \frac{1}{\mu}.$$
It remains to show that $\lim_{t \to \infty} \frac{m(t)}{t} \le \frac{1}{\mu}.$

If the interarrival times X_1, X_2, \ldots are bounded by a constant M, then the residual life Y(t) is also bounded by M. Hence,

$$\lim_{t \to \infty} \frac{m(t)}{t} \leq \lim_{t \to \infty} \frac{1}{\mu} - \frac{1}{t} + \frac{M}{t\mu} = \frac{1}{\mu}$$

The Elementary Renewal Theorem for renewal process with **bounded interarrival times** is proved.

Proof of the Elementary Renewal Theorem (Cont'd)

In general, if the interarrival times X_1, X_2, \ldots are not bounded, we fix a constant M and define a new renewal process $N_M(t)$ with the truncated interarrival times

 $\min(X_1, M), \min(X_2, M), \ldots, \min(X_n, M), \ldots$

Because min $(X_i, M) \leq X_i$ for all *i*, it follows that $N_M(t) \geq N(t)$ for all *t*.

$$\lim_{t\to\infty}\frac{m(t)}{t} = \lim_{t\to\infty}\frac{\mathbb{E}[N(t)]}{t} \le \lim_{t\to\infty}\frac{\mathbb{E}[N_M(t)]}{t} = \frac{1}{\mathbb{E}[\min(X_1, M)]}$$

by the Elementary Renewal Theorem with bounded interarrival times. Note the inequality above is valid for all M > 0. Letting $M \to \infty$ yields

$$\lim_{t\to\infty}\frac{m(t)}{t}\leq \frac{1}{\mu}.$$

Here we use the fact that $\mathbb{E}[\min(X_1, M)] \to \mathbb{E}[X_1] = \mu$ as $M \to \infty$. Lecture 15 - 12

Example 7.6 (M/G/1 with no Queue)

- Single-server bank
- Potential customers arrive at a Poisson rate λ
- Customers enter the bank only if the server is free
- Service times are i.i.d. with mean μ_{G} , indep. of the arrival
- Let N(t) = number of customers entry the bank by time t and those who arrive finding the server busy and walk away don't count. Is {N(t) : t ≥ 0} a (delayed) renewal process?

Ans. An interarrival time $T_i = G_i + W_i$ where

 $G_i =$ service time, i.i.d., w/ mean μ_G

 W_i = waiting time until the next customer arrives after the previous one

As potential customers arrive following a Poisson process, by the memoryless property, W_i 's are i.i.d. $Exp(\lambda)$.

The interarrival times $\{T_i\} = \{G_i + W_i\}$ are i.i.d. The events of customers entering constitutes a renewal process

Example 7.6 (M/G/1 with no Queue)

- Q: What is the rate at which customers enter the bank?
- As $\mathbb{E}[T_i] = \mathbb{E}[G_i] + \mathbb{E}[W_i] = \mu_G + \frac{1}{\lambda}$, by the Elementary Renewal Theorem, the rate is

$$\frac{1}{\mathbb{E}[T_i]} = \frac{1}{\mu_G + \frac{1}{\lambda}} = \frac{\lambda}{\lambda\mu_G + 1}$$

- **Q**: What is the proportion of potential customers that are lost?
 - As potential customers arrive at rate λ , and customers enter at the rate $\frac{\lambda}{\lambda\mu_G + 1}$, the proportion that actually enter the bank is

$$rac{\lambda/(\lambda\mu_{G}+1)}{\lambda}=rac{1}{\lambda\mu_{G}+1}$$

So the proportion that is lost is $1 - \frac{1}{\lambda \mu_G + 1} = \frac{\lambda \mu_G}{\lambda \mu_G + 1}.$