

The Lecture Contains:

- ☰ Renewal Theory
- ☰ What is Convolution
- ☰ Proposition
- ☰ Proof of Proposition

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Renewal Theory

We know that for a Poisson process, the inter arrival time (a some sort of **counting process** which counts the inter arrival time in this case) are **independent** and **identically distributed (i.i.d)** random variables (r.v.'s) where the underlying distribution is exponential in nature. Consider the **counting process** has an underlying distribution and let it be denoted by any arbitrary distribution say, $F_X(x)$. Then this general counting process is termed as a **renewal process**.

Before going into details of a **renewal process**, let us formalize the definition of a counting process. Given any arbitrary distribution, let $\{X_n; n = 1, 2, \dots\}$ be a sequence of non-negative independent random variables (r.v.'s) (example being the time between the $(n-1)^{th}$ and the n^{th} event) with a common distribution, $F_X(x)$ such that $F_X(0) = P\{X_n = 0\} < 1$. Let the **mean time between successive events** be denoted by $\mu = E(X_n) = \sum_{x=0}^{\infty} \{x \times P(x)\}$. Now $X_n \geq 0$ and $F_X(0) = P\{X_n = 0\} < 1$, will ensure that $0 < \mu \leq \infty$.

Furthermore let us denote $S_n = \sum_{i=1}^n X_i = (X_1 + X_2 + \dots + X_n)$, where $n \geq 1$, as the sum of the **inter arrival times** of n number of such events or better still as the time of the n^{th} event, with the added condition that $S_0 = 0$, Figure 4.1.

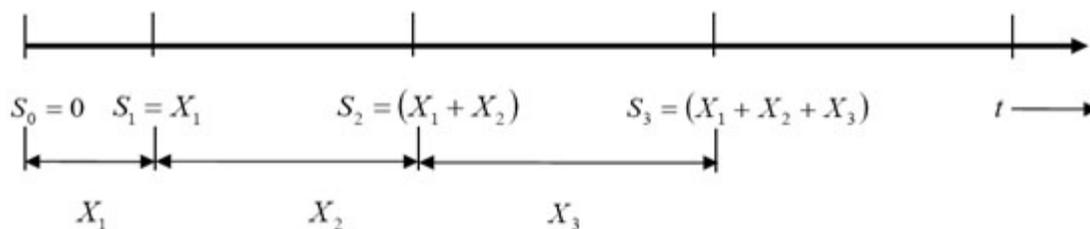


Figure 4.2: Schematic diagram to illustrate the concept of a renewal process using interarrival times and sum of interarrival times

Remember, the number of events by time t will equal the largest value of n for which the n^{th} event occurs before or at time t . We already have $N(t)$, as the number of events by time t which is given by $N(t) = \sup\{n : S_n \leq t\}$, and this $\{N(t), t \geq 0\}$ is what we call a **renewal process**.

Note

- Since the inter arrival times are *i.i.d*, hence at each renewal or event the process probability starts over again.
- There are finite numbers of renewals which can occur in a finite time, as by the **strong law of large numbers (SLLN)** we can show that

$$P\left\{\lim_{n \rightarrow \infty} \left(\frac{S_n}{n}\right) = \mu\right\} = P\left\{\lim_{n \rightarrow \infty} \left(\frac{X_1 + \dots + X_n}{n}\right) = \mu\right\} = 1. \text{ But since } 0 < \mu \leq \infty, \text{ hence } S_n \text{ must}$$

go towards infinity as n goes to infinity. Thus, S_n can be less than or equal to t for at most a finite number of values of n , hence $N(t) = \sup\{n: S_n \leq t\}$ must be finite and one can write $N(t) = \max\{n: S_n \leq t\}$

For a better understanding of the concept of **renewal theory** it is important that we find the distribution of $N(t)$, but before that one must note the important relationship that the number of renewals by time t is greater than or equal to n iff the n^{th} renewal occurs before or at time t . Hence we need to check the following theorem.

Theorem 4.1

$$N(t) \geq n \Leftrightarrow S_n \leq t$$

Proof 4.1

$$N(t) \geq n \Leftrightarrow S_n \leq t$$

$$\text{i.e., } P\{N(t) = n\} = P\{N(t) \geq n\} - P\{N(t) \geq n+1\}$$

$$= P\{S_n \leq t\} - P\{S_{n+1} \leq t\}$$

Now $X_i, i \geq 1$ are *i.i.d*, hence it follows that $S_n = \sum_{i=1}^n X_i$ is distributed F_n which is the n -fold convolution of F itself which is the distribution of $X_i, i \geq 1$.

Hence

$$P\{N(t) = n\} = F_n(t) - F_{n+1}(t)$$



What is convolution

In general functional analysis convolution means that if we have two different functions, f and g

producing a third function, then we denote it by $(f * g)(x) = \int_{-\infty}^{+\infty} f(\tau)g(x-\tau)d\tau = \int_{-\infty}^{+\infty} f(x-\tau)g(\tau)d\tau$,

where the operator denoted by $*$ is what is termed as convolution. In general the properties followed by convolution, considering there are two functions, f and g , are:

1. **Commutative property**, i.e., $f * g = g * f$
2. **Associative property**, i.e., $f * (g * h) = (f * g) * h$
3. **Distributive property**, i.e., $f * (g + h) = (f * g) + (f * h)$
4. **Associative property with scalar multiplication**, i.e., $\alpha(f * g) = (\alpha f) * g = f * (\alpha g)$

Let us now denote $m(t) = E[N(t)] = \sum_{n=1}^{\infty} n \times P\{N(t) = n\}$ as the **renewal function** and one of our main

focus in **renewal theory** is to determine the properties of this **renewal function**. A natural question which arises is how does one understand the importance of this renewal function.

Proposition 4.2

$$m(t) = E[N(t)] = \sum_{n=1}^{\infty} n \times P\{N(t) = n\} = \sum_{n=1}^{\infty} F_n(t)$$

Proof of Proposition 4.2

We can write

$$N(t) = \sum_{n=1}^{\infty} I_n, \text{ where } I_n = \begin{cases} 1 & \text{if the } n^{\text{th}} \text{ renewal occurred in } [0, t] \\ 0 & \text{otherwise} \end{cases}$$

Hence

$$\begin{aligned} E[N(t)] &= E\left[\sum_{n=1}^{\infty} I_n\right] = \sum_{n=1}^{\infty} E(I_n) \\ &= \sum_{n=1}^{\infty} \{1 \times P(I_n = 1)\} + \{0 \times P(I_n = 0)\} \\ &= \sum_{n=1}^{\infty} P(I_n = 1) \\ &= \sum_{n=1}^{\infty} P(S_n \leq t) = \sum_{n=1}^{\infty} F_n(t) \end{aligned}$$