



Chapter 5 Poisson Processes

Dong Yan
EI.HUST.

Chapter 5: Poisson Processes



OUTLINE

5.1 Poisson Processes (2.1,2.2)

5.2 Generalization of the Poisson Processes
(2.3, 2.4)

5.3 Filtered Poisson Processes (2.5)

5.4 Two-Dimensional and Marked Poisson
Processes (2.6)

5.5 Poisson Arrival See Time Averages (2.7)



5.3 Filtered Poisson Processes

◆ Compound Poisson Processes:
$$X(t) = \sum_{n=1}^{N(t)} Y_n$$

Sometimes $\{Y_n\}$ cannot be used directly but the function of Y_n .

Definition of filtered Poisson Processes:

A stochastic process $\{X(t), t \geq 0\}$ is called a filtered Poisson process if

$$X(t) = \sum_{n=1}^{N(t)} w(t, S_n, Y_n)$$

where $N(t)$ is a **Poisson process** with intensity ν and $\{Y_n\}$ are **identical independent distribution random variables**. The process $N(t)$ and the sequence $\{Y_n\}$ are **independent**. $\{S_n\}$ are arrival times. $w()$ is the response function defined by the three arguments.



5.3 Filtered Poisson Processes

◆ If $w(t, S_n, Y_n) = \begin{cases} 1 & t \geq 0 \\ 0 & t < 0 \end{cases}$ $X(t) = \sum_{n=1}^{N(t)} w(t, S_n, Y_n)$

then $X(t) = N(t)$, $X(t)$ is a Poisson process.

◆ If $w(t, S_n, Y_n) = \begin{cases} Y_n & t \geq 0 \\ 0 & t < 0 \end{cases}$

then $X(t) = \sum_{n=1}^{N(t)} Y_n$, $X(t)$ is a compound Poisson process.

◆ A frequently used function assumes the following form:

$$w(t, \tau, y) = w(t - \tau, y)$$



5.3 Filtered Poisson Processes

Example 1: (example 2.2.3 in textbook)

A cable TV company collects \$1/unit time from each subscriber. Subscribers sign up in accordance with a Poisson process with rate λ . What is the expected total revenue received in $(0, t]$?

Let S_i denote the arrival time of the i th customer.

The revenue generated by this customer in $(0, t]$ is $t - S_i$.

The total revenue received in $(0, t]$ is $X(t) = \sum_{i=1}^{N(t)} (t - S_i)$

$Y(t)$ is a filtered Poisson process, if define

$$w(t, S_i, Y_i) = \begin{cases} t - S_i & t \geq S_i \\ 0 & t < S_i \end{cases}$$



5.3 Filtered Poisson Processes

Probability generating function of $X(t)$:

Let $\{U_i\}$ be i.i.d. random variables with uniform distribution on $(0, t)$.

$$\begin{aligned} E[z^{X(t)} \mid N(t) = n] &= E[z^{\sum_{i=1}^{N(t)} w(t, S_i, Y_i)} \mid N(t) = n] \\ &= E[z^{\sum_{i=1}^n w(t, U_i, Y_i)}] \quad U_i, Y_i \text{ are i.i.d. random variables respectively} \\ &= (E[z^{w(t, U, Y)}])^n \\ E[z^{X(t)}] &= \sum_{n=0}^{\infty} e^{-\lambda t} \frac{(\lambda t)^n}{n!} (E[z^{w(t, U, Y)}])^n \\ &= \exp(-\lambda t) \exp\{\lambda t E[z^{w(t, U, Y)}]\} \\ &= \exp\{\lambda t \{E[z^{w(t, U, Y)}] - 1\}\} \end{aligned}$$



5.3 Filtered Poisson Processes

Mean function of $X(t)$:

$$\begin{aligned} E[X(t)] &= \frac{dE[z^{X(t)}]}{dz} \Big|_{z=1} = \frac{d\{\lambda t\{E[z^{w(t,U,Y)}] - 1\}\}}{dz} \Big|_{z=1} \\ &= \lambda t \frac{dE[z^{w(t,U,Y)}]}{dz} \Big|_{z=1} = \lambda t E\left[\frac{dz^{w(t,U,Y)}}{dz} \Big|_{z=1}\right] \\ &= \lambda t E[w(t,U,Y) z^{w(t,U,Y)-1} \Big|_{z=1}] \end{aligned}$$

$$E[X(t)] = \lambda t E[w(t,U,Y)]$$

Variance function of $X(t)$: $Var[X(t)] = \lambda t E[w^2(t,U,Y)]$

Mean of compound Poisson process: $E[X(t)] = \lambda t E[Y]$

Variance of compound Poisson process: $Var[X(t)] = \lambda t E[Y^2]$

Note: $w(t,U,Y)$ includes two random variables: U, Y



5.3 Filtered Poisson Processes

Example 1: (example 2.2.3 in textbook)

A cable TV company collects \$1/unit time from each subscriber. Subscribers sign up in accordance with a Poisson process with rate λ . What is the expected total revenue received in $(0, t]$?

$$X(t) = \sum_{i=1}^{N(t)} (t - S_i)$$

$X(t)$ is a filtered Poisson process, if define $w(t, S_i, Y_i) = \begin{cases} t - S_i & t \geq S_i \\ 0 & t < S_i \end{cases}$

Obtain: $E[X(t)] = \lambda t E[w(t, U, Y)]$

$$= \lambda t E[w(t, U)] = \lambda t \int_0^t (t - u) \frac{1}{t} du$$

$$= \lambda \left[tu \Big|_0^t - \frac{1}{2} u^2 \Big|_0^t \right] = \lambda \left[t^2 - \frac{1}{2} t^2 \right]$$

$$E[X(t)] = \frac{1}{2} \lambda t^2$$

Chapter 5: Poisson Processes



OUTLINE

5.1 Poisson Processes (2.1,2.2)

5.2 Generalization of the Poisson Processes
(2.3, 2.4)

5.3 Filtered Poisson Processes (2.5)

5.4 Two-Dimensional and Marked Poisson
Processes (2.6)

5.5 Poisson Arrival See Time Averages (2.7)

5.4 Two-Dimensional and Marked Poisson Processes



1. Two-Dimensional Poisson Processes

S denotes a two-dimensional plane. Let A be a subset of plane S . Points are scattered randomly over S .

Let $N(A)$ be the number of points in A .

$N(A)$ is called a point process in S .

Stochastic process $\{N(A), A \subset S\}$ is a *two-dimensional Poisson process* if

- i) $N(A)$ follows a Poisson distribution with mean $\lambda |A|$.
- ii) the numbers of points occurring in disjoint subsets of S are mutually independent.

where $|A|$ denote the size of the set A .

5.4 Two-Dimensional and Marked Poisson Processes



Two-dimensional nonhomogeneous Poisson processes:

Let $\lambda(x, y)$ be the intensity function of the Point process $N(A)$. The process is a two-dimensional nonhomogeneous Poisson process if

- i) $N(A)$ follows a Poisson distribution with mean $\iint \lambda(x, y) dx dy$
- ii) the numbers of points occurring in disjoint subsets of S are mutually independent.

n -dimensional Poisson processes

n -dimensional nonhomogeneous Poisson processes

Stars in space.

5.4 Two-Dimensional and Marked Poisson Processes



2. Marked Poisson processes

$N(t)$ is a Poisson process with mean λ . $\{S_n\}$ are the arrival times of $N(t)$. Let $\{Y_n\}$ be i.i.d. random variables with a common distribution $G(y)$ (or $g(y)$), where Y_n is associated with the n th arrival of $N(t)$. $N(t)$ and $\{Y_n\}$ are independent.

The stochastic process (S_n, Y_n) defined in the (t, y) plane is called a *marked Poisson process*.

$$\therefore P\{Y \in (y, y+h)\} = g(y)h + o(h)$$

The process (S_n, Y_n) (or $N(A)$) is a two-dimensional nonhomogeneous Poisson process with intensity function $\lambda(t, y) = \lambda g(y)$.

5.4 Two-Dimensional and Marked Poisson Processes



Mean of $N(A)$:

$$E[N(A)] = \iint_A \lambda g(y) dy dt$$

$$E[N(A)] = \iint_A \lambda dG(y) dt$$

Example:

G assumes the form of a multinomial distribution with

$$P\{Y = i\} = p_i, \quad i = 1, \dots, n; \quad p_1 + p_2 + \dots + p_n = 1$$

Then the Poisson process $N(A)$ is decomposed into n independent Poisson streams, each with density λp_i respectively.

Chapter 5: Poisson Processes



OUTLINE

5.1 Poisson Processes (2.1,2.2)

5.2 Generalization of the Poisson Processes
(2.3, 2.4)

5.3 Filtered Poisson Processes (2.5)

5.4 Two-Dimensional and Marked Poisson
Processes (2.6)

5.5 Poisson Arrival See Time Averages
(2.7)



5.5 Poisson Arrival See Time Averages

- ◆ Important in queuing modeling
 - ◆ In a service system, customers arrival process is Poisson.
 - ◆ $X(s)$ represents the number of customers in the system at time $0 < s < t$.
 - ◆ So $X(t)$ is a stochastic process with a discrete state space $S = \{0, 1, \dots\}$.
 - ◆ B is a subset of S .
1. The ratio of the sojourn time $X(s)$ ($0 < s < t$) in B and the total time t .
 2. The fraction of Poisson arrivals who find $X(s)$ ($0 < s < t$) in B .

5.5 Poisson Arrival See Time Averages



- ◆ the *lack of anticipation assumption* (LAA) : for each $t \geq 0$, the arrival process $\{N(t+u) - N(t), u \geq 0\}$ is independent of $\{N(s), 0 \leq s \leq t\}$ and $\{X(s), 0 \leq s \leq t\}$
- ◆ Poisson processes satisfy LAA.



5.5 Poisson Arrival See Time Averages

1. The sojourn time $X(s)$ ($0 < s < t$) in B

◆ An indicator random variable $U(t)$:

$$U(t) = \begin{cases} 1 & \text{if } X(t) \in B \\ 0 & \text{otherwise} \end{cases}$$

Assume that $U(t)$ is left-continuous function.

$\int_0^t u(s) ds$ is the time the process $X(s)$ is in the set B in $[0, t]$.

◆ The fraction of time the process $X(s)$ is in the set B in $[0, t]$:

$$V(t) = \frac{\int_0^t U(s) ds}{t}$$



5.5 Poisson Arrival See Time Averages

2. The fraction of Poisson arrivals who find $X(s)$ ($0 < s < t$) in B.
(who sees the system in state set B)

◆ The number of arrivals in $(0, t]$ who find $X(s)$ ($0 < s < t$) in B:

$$Y(t) = \int_0^t U(s) dN(s)$$

◆ The fraction of Poisson arrivals who find $X(s)$ ($0 < s < t$) in B:

$$Z(t) = \frac{Y(t)}{N(t)}$$

◆ Under LAA, $P\{\lim_{n \rightarrow \infty} Z_n(t) = V(t)\} = 1$

The fraction of arrivals that sees the system in state set B is
equal to the fraction of the time process is in that state set.

5.5 Poisson Arrival See Time Averages



$$E\{Y(t)\} = \lim_{n \rightarrow \infty} E\{Y_n(t)\} = \lambda t E\{V(t)\} = \lambda E\left\{\int_0^t U(s) ds\right\}$$

The expected number of arrivals who see the system in state set B in $[0, t]$ is **equal to** the arrival rate multiplied by the length of time the system has been in B in $[0, t]$.



End of Chapter 5

