

The Counting Process of Which the Intensity Function Depends on States

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Abstract

In this paper we are concerned with the counting processes with intensity function $g_n(t)$, where $g_n(t)$ not only depends on t but n . It is shown that under certain conditions the number of events in $[0, t]$ follows a generalized Poisson distribution. A counting process is also provided such that $g_i(t) \neq g_j(t)$ for $i \neq j$ and the number of events in $[0, t]$ has a transformed geometric distribution.

1. Introduction

The Poisson distribution has been generalized in many ways. Rao and Rubin (1964), Chon (1960) and Singh (1966) introduced a generalized Poisson distribution. Consul and Jain (1970) presented a new generalized Poisson distribution with two parameter (θ, λ) . Recently, Consul (1988) investigated some models leading to the generalized Poisson distribution and Consul (1989) introduced a generalized Poisson process for which the number of events in interval $[0, t]$ is generalized Poisson distributed with parameters $(\theta t, \lambda t)$.

In Poisson processes, the intensity function is $g_n(t) = \lambda$ for all $n = 1, 2, \dots$, and the intensity function of nonhomogeneous Poisson process is $g_n(t) = \lambda(t)$ for all $n = 1, 2, \dots$. Therefore, the Poisson process and nonhomogeneous Poisson process have intensity function which does not depend on state n . Consul (1989) proposed a generalized Poisson process of which that intensity function depends on state n .

In this paper, we provide some conditions under which the number of events in $[0, t]$ follows a generalized Poisson distribution, and show that the definition of a generalized Poisson process defined by Consul(1989) is not well-defined. In section 2, we propose a counting process such that the intensity function is $g_i(t) \neq g_j(t)$ for $i \neq j$.

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Let $\{N(t) \mid t \geq 0\}$ be a counting process having jump magnitude 1. Then counting process $\{N(t) \mid t \geq 0\}$ is satisfies

$$P\{N(t+h) - N(t) \geq 2\} = o(h). \tag{1.1}$$

Suppose that $P\{N(t+h) - N(t) = 1 \mid N(t) = n\} = g_n(t, h)$ and $g_n(t, h)$ is a polynomial function with respect to h in which the constant term is zero;

$$g_n(t, h) = g_n(t) h + g_n(t) h^2 + \dots = g_n(t) h + o(h).$$

Then

$$P\{N(t+h) - N(t) = 1 \mid N(t) = n\} = g_n(t) h + o(h) \tag{1.2}$$

and

$$P\{N(t+h) - N(t) = 0 \mid N(t) = n\} = 1 - g_n(t) h + o(h) \dots$$

Now $g_n(t)$ is called the *intensity function* of counting process $\{N(t) \mid t \geq 0\}$.

By Equations (1.1) and (1.2),

$$\begin{aligned} P\{N(t+h) = n\} &= P\{N(t+h) - N(t) = 0 \mid N(t) = n\} P\{N(t) = n\} \\ &\quad + P\{N(t+h) - N(t) = 1 \mid N(t) = n-1\} P\{N(t) = n-1\} \\ &\quad + \sum_{i=2}^{\infty} P\{N(t+h) - N(t) = i \mid N(t) = n-i\} P\{N(t) = n-i\} \\ &= \{1 - g_n(t) h\} P\{N(t) = n\} + g_{n-1}(t) h P\{N(t) = n-1\} + o(h). \end{aligned}$$

Hence,

$$\begin{aligned} \frac{P\{N(t+h) = n\} - P\{N(t) = n\}}{h} &= g_n(t) P\{N(t) = n\} \\ &\quad + g_{n-1}(t) P\{N(t) = n-1\} + \frac{o(h)}{h}. \end{aligned}$$

Letting $h \rightarrow 0$, we obtain the differential equation

$$\frac{dP\{N(t) = n\}}{dt} = g_n(t) P\{N(t) = n\} + g_{n-1}(t) P\{N(t) = n-1\}.$$

The solution of the above equation is

$$P\{N(t) = n\} = e^{-\int g_n(t) dt} \int g_{n-1}(t) P\{N(t) = n-1\} e^{\int g_n(t) dt} dt + k_n e^{-\int g_n(t) dt}. \tag{1.3}$$

If the counting process is a Poisson or nonhomogeneous Poisson, we know that $k_0 = 1$ and $k_n = 0$ ($n \geq 1$). The constants $\{k_0, k_1, k_2, \dots\}$ is called the *integral constants* of the corresponding counting process. From (1.3) we can see that the distribution of $N(t)$, the

number of events in interval $[0, t]$, depends on $g_n(t)$ and $g_{n-1}(t)$. It is well-known that $g_n(t) = \lambda$ for all n is a necessary condition for the counting process to be a Poisson process with rate λ . Similarly, $g_n(t) = \lambda(t)$ for all n is necessary for the counting process to be a nonhomogeneous Poisson process with intensity function $\lambda(t)$. As can be seen, $g_n(t)$ does not depend on n for the cases of Poisson and nonhomogeneous Poisson processes. In general, however, $g_n(t)$ depends on n in generalized Poisson processes.

2. P-process

Let $\int_* \mathcal{A}(t) dt = \int \mathcal{A}(t) dt - C$, where C is a integral constant of $\mathcal{A}(t)$.

The function $\mathcal{A}(t)$ is said to be a *t-zero function* if $\left[\int_* \mathcal{A}(t) dt \right]_{t=0} = 0$.

[Definition 1] The counting process $\{N(t) \mid t \geq 0\}$ is said to be a *polynomial process* (*P-process*) with intensity function $g_n(t)$ if

- (i) $N(0) = 0$,
- (ii) $P\{N(t+h) - N(t) = 1 \mid N(t) = n\} = g_n(t)h + o(h)$

where $-\infty < \left[\int_* g_n(t) dt \right]_{t=0} < \infty$,

- (iii) $P\{N(t+h) - N(t) \geq 2 \mid N(t) = n\} = o(h)$ for each $n = 0, 1, 2, \dots$.

If $g_n(t) = \lambda$ for each $n = 0, 1, 2, \dots$, then the P-process is a Poisson process with rate λ , and if $g_n(t) = \lambda(t)$ for each $n = 0, 1, 2, \dots$, then the P-process is a nonhomogeneous Poisson process with intensity function $\lambda(t)$.

Let $P_n(t) = P\{N(t) = n\}$. Then, from the definition and (1.3), we obtain that

$$P_0(t) = k_0 \exp\left(- \int_* g_0(t) dt\right),$$

and for $n \geq 1$,

$$P_n(t) = \exp\left(- \int_* g_n(t) dt\right) \left[\int_* g_{n-1}(t) P_{n-1}(t) \exp\left(\int_* g_n(t) dt\right) dt \right] + k_n \exp\left(- \int_* g_n(t) dt\right),$$

where k_0, k_1, \dots are constants.

[Proposition 1] If $\{N(t) \mid t \geq 0\}$ is a P-process with intensity function $g_n(t)$, then

$$(1) \quad g_0(t) = -\frac{P_0'(t)}{P_0(t)}.$$

$$(2) \quad g_n(t) = \frac{g_{n-1}(t)P_{n-1}(t)}{P_n(t)} - \frac{P_n'(t)}{P_n(t)}, \quad n \geq 1.$$

By the boundary condition $P_0(0) = 1$ and $P_n(0) = 0$ ($n \geq 1$), we obtain the following :

[Theorem 2] Let $\{N(t) \mid t \geq 0\}$ be a P-process with intensity function $g_n(t)$. Then,

(1) $g_0(t)$ is a t-zero function if and only if $k_0 = 1$.

(2) $g_{n-1}(t)P_{n-1}(t) \exp\left(\int g_n(t) dt\right)$ ($n \geq 1$) is a t-zero function if and only if $k_n = 0$.

(Proof) (1) By the boundary condition $P_0(0) = 1$,

$$\begin{aligned} g_0(t) \text{ be a t-zero function} &\Rightarrow \left[-\int_* g_0(t) dt\right]_{t=0} = 0 \\ &\Rightarrow P_0(0) = k_0 \exp(0) = 1 \\ &\Rightarrow k_0 = 1 \\ &\Rightarrow P_0(0) = \left[\exp\left(\int_* g_0(t) dt\right)\right]_{t=0} = 1 \\ &\Rightarrow g_0(t) \text{ be a t-zero function.} \end{aligned}$$

(2) By the boundary condition $P_n(0) = 0$ ($n \geq 1$),

$g_{n-1}(t)P_{n-1}(t) \exp\left(\int g_n(t) dt\right)$ ($n \geq 1$) is a t-zero function

$$\begin{aligned} &\Rightarrow k_n \exp\left(-\int_* g_n(t) dt\right) = P_n(0) = 0 \\ &\Rightarrow k_n = 0 \\ &\Rightarrow P_n(0) = \left[\exp\left(-\int_* g_n(t) dt\right) \int_* g_{n-1}(t)P_{n-1}(t) \exp\left(\int_* g_n(t) dt\right) dt\right]_{t=0} = 0 \\ &\Rightarrow \left[\int_* g_{n-1}(t)P_{n-1}(t) \exp\left(\int_* g_n(t) dt\right) dt\right]_{t=0} = 0 \\ &\Rightarrow g_{n-1}(t)P_{n-1}(t) \exp\left(\int g_n(t) dt\right) \text{ ($n \geq 1$) is a t-zero function.} \quad \square \end{aligned}$$

[Definition 2] The P-process $\{N(t) \mid t \geq 0\}$ is called a *strongly P-process* if

$$k_0 = 1 \text{ and } k_n = 0 (n \geq 1).$$

[Proposition 3] If $\{N(t) \mid t \geq 0\}$ is a strongly P-process with intensity function $g_n(t)$, then

$$g_0(0) = P_1(0) \text{ and } P_n(0) = 0 (n \geq 2).$$

(Proof) For $n \geq 1$,

$$k_n = \frac{g_{n-1}(0)P_{n-1}(0) - P_n(0)}{g_n(0)} \left[\exp\left(\int_0^* g_n(t) dt\right) \right]_{t=0} - \left[\int_0^* g_{n-1}(t) P_{n-1}(t) \exp\left(\int_0^* g_n(t) dt\right) dt \right]_{t=0}.$$

Since $\{N(t) \mid t \geq 0\}$ is a strongly P-process, $k_n = 0$. By [Theorem 2],

$$\left[\int_0^* g_{n-1}(t) P_{n-1}(t) \exp\left(\int_0^* g_n(t) dt\right) dt \right]_{t=0} = 0.$$

Hence,

$$\frac{g_{n-1}(0)P_{n-1}(0) - P_n(0)}{g_n(0)} = 0,$$

which implies that

$$g_{n-1}(0)P_{n-1}(0) - P_n(0) = 0.$$

Taking $n = 1$. By the boundary condition $P_0(0) = 1$,

$$g_0(0) = P_1(0).$$

If $n \geq 2$, $P_n(0) = 0$ and thus $P_n(0) = 0$. \square

The Poisson process is a strongly P-process because $g_0(t) = \lambda$ is a t-zero function, and

$$g_{n-1}(t)P_{n-1}(t) \exp\left(\int_0^* g_n(t) dt\right) = \frac{\lambda^n t^{n-1}}{(n-1)!}$$

is a t-zero function. For the case of nonhomogeneous Poisson processes, if intensity function $\lambda(t)$ is a t-zero function, then the nonhomogeneous Poisson process is a strongly P-process.

The following example shows that there exists a strongly P-process such that $g_i(t) \neq g_j(t)$ for some i, j .

[Example 1] Let the counting process $\{N(t) \mid t \geq 0\}$ satisfy

- (1) $N(0) = 0$,
- (2) $P\{N(t+h) - N(t) = 1 \mid N(t) = 0\} = \left(\theta - \frac{\theta\lambda}{1 + \theta\lambda t}\right)h + o(h)$,
- (3) $P\{N(t+h) - N(t) = 1 \mid N(t) = 1\} = \left(\frac{\theta}{1 - \lambda}\right)h + o(h)$ where $0 \leq \lambda < 1$,
- (4) $P\{N(t+h) - N(t) = 1 \mid N(t) = n\} = \lambda h + o(h)$, $n = 2, 3, 4, \dots$,
- (5) $P\{N(t+h) - N(t) \geq 2 \mid N(t) = n\} = o(h)$ for each $n = 0, 1, 2, \dots$.

Since $g_0(t) = \theta - \frac{\theta\lambda}{1 + \theta\lambda t}$ is a t -zero function, $k_0 = 1$ by Theorem 2. and

$$P_0(t) = e^{-\theta t}(1 + \theta\lambda t).$$

Since

$$g_0(t)P_0(t) \exp\left(\int_* g_1(t) dt\right) = (\theta - \theta t + \theta^2\lambda t) \exp\left(\frac{\theta\lambda t}{1 - \lambda}\right)$$

is a t -zero function, by Theorem 2, $k_1 = 0$. Thus we obtain

$$P_0(t) = \theta t e^{-\theta t}(1 - \lambda).$$

For $n \geq 2$,

$$g_{n-1}(t)P_{n-1}(t) \exp\left(\int_* g_n(t) dt\right) = \frac{\theta^n}{(n-1)!} t^{n-1}$$

is a t -zero function. Also, by Theorem 2, $k_n = 0$ and we obtain

$$P_n(t) = \frac{(\theta t)^n}{n!} e^{-\theta t}.$$

Thus counting process $\{N(t) \mid t \geq 0\}$ is a strongly P-process and the distribution of number of events in the interval $[0, t]$ is

$$P\{N(t) = n\} = \begin{cases} e^{-\theta t}(1 + \theta\lambda t), & n = 0 \\ \theta t e^{-\theta t}(1 - \lambda), & n = 1 \\ \frac{(\theta t)^n}{n!} e^{-\theta t}, & n = 2, 3, 4, \dots \end{cases}$$

The above distribution is a generalized Poisson distribution with parameters $(\theta t, \lambda)$ defined by Rao and Rubin (1964). For convenience, this strongly P-process is called to be a $(0,1)$ -generalized Poisson process. \square

The following example is a P-process but not strongly P-process.

[Example 2] Let counting process $\{N(t) \mid t \geq 0\}$ satisfy

- (1) $N(0) = 0,$
- (2) $P\{N(t+h) - N(t) = 1 \mid N(t) = n\} = \theta h + o(h), \quad n = 0, 3, 4, 5, \dots,$
- (3) $P\{N(t+h) - N(t) = 1 \mid N(t) = 1\} = \left(\theta - \frac{2\theta\lambda}{2 + \theta\lambda t}\right)h + o(h),$
- (4) $P\{N(t+h) - N(t) = 1 \mid N(t) = 2\} = \left(\frac{\theta}{1 - \lambda}\right)h + o(h)$ where $0 \leq \lambda < 1,$
- (5) $P\{N(t+h) - N(t) \geq 2 \mid N(t) = n\} = o(h)$ for each $n = 0, 1, 2, \dots$

Since $g_n(t)$ is a t-zero function except $n = 1$ and

$$\left[\int_* g_1(t) dt \right]_{t=0} = \left[\int_* \left(\theta - \frac{2\theta\lambda}{2 + \theta\lambda t} \right) dt \right]_{t=0} = -2 \ln 2 < \infty,$$

we know that counting process $\{N(t) \mid t \geq 0\}$ is a P-process.

By Theorem 2, $g_0(t) = \theta$ is a t-zero function imply $k_0 = 1$. However,

$$\begin{aligned} g_0(t) P_0(t) \exp\left(\int g_1(t) dt\right) &= \theta \exp(-\theta t) \exp\left(\int \left(\theta - \frac{2\theta\lambda}{2 + \theta\lambda t}\right)\right) \\ &= -\frac{\theta}{(2 + \theta\lambda t)^2} \end{aligned}$$

is not a t-zero function. Thus $\{N(t) \mid t \geq 0\}$ is not a strongly P-process. In this case we get the integral constants $k_1 = 1/2\lambda$ and $0 = k_2 = k_3 = \dots$

The distribution of number of events in the interval $[0, t]$ is the following

$$P\{N(t) = n\} = \begin{cases} e^{-\theta t}, & n = 0 \\ \theta t e^{-\theta t} \left(1 + \frac{\theta\lambda t}{2}\right), & n = 1 \\ \frac{(\theta t)^2}{2!} e^{-\theta t} (1 - \lambda), & n = 2 \\ \frac{(\theta t)^n}{n!} e^{-\theta t}, & n = 3, 4, \dots \end{cases}$$

For convenience, we call the above distribution by *(1,2)-generalized Poisson distribution* with parameters $(\theta t, \lambda)$ and the associated P-process *(1,2)-generalized Poisson process* with parameters $(\theta t, \lambda)$. \square

Consul and Jain (1973) presented a new generalized Poisson distribution with two parameters θ and λ . The definition of generalized Poisson distribution is as follows :

[Definition] (Consul and Jain) The discrete random variable X is said to be a generalized Poisson distribution with parameters (θ, λ) if a probability mass function of X is

$$P_X(\theta, \lambda) = \begin{cases} \frac{\theta(\theta + x\lambda)^{x-1} e^{-\theta - x\lambda}}{x!}, & \text{if } x = 0, 1, 2, \dots \\ 0, & \text{for } x > m, \text{ when } \lambda < 0 \text{ and otherwise} \end{cases}$$

where $\theta > 0$, $\max(-1, -\theta/m) < \lambda \leq 1$ and $m(\geq 4)$ is the largest positive for which $\theta + m\lambda > 0$ when λ is negative.

Consul (1989) introduced a new generalized Poisson process where the number of changes $N(t)$ in an interval $[0, t]$ has a generalized Poisson distribution with parameters $(\theta t, \lambda t)$. He made a suggestion to adopt four axioms of new generalized Poisson process. The four axioms are the following :

(Axiom 1) $N(0) = 0$.

(Axiom 2) For any $t > 0$, $0 < P\{N(t) > 0\} < 1$.

(Axiom 3) Direct transitions from state i are possible only to state $i+1$ and not to the higher states; that is, for any $t > 0$, in sufficiently small intervals of duration h , at most one event can occur.

(Axiom 4) $P\{N(t+h) - N(t) = 0 \mid N(t) = n\} = 1 - (\theta + n\lambda)h + o(h)$

$$P\{N(t+h) - N(t) = 1 \mid N(t) = n-1\} = \frac{(\theta + (n-1)\lambda)he^{-\lambda t}}{(1 - \lambda(\theta + n\lambda))^{n-1}} + o(h).$$

From Axioms, we know that a new generalized Poisson process is a P-process. Consul insisted that the integral constants of new generalized Poisson process are $k_0 = 1$ and $k_n = 0 (n \geq 1)$, and the distribution of number of events in interval $[0, t]$ is a new generalized Poisson with parameters $(\theta t, \lambda t)$. Therefore if Consul's results are correct, the new generalized Poisson process is a strongly P-process. But, under the above Axioms, we cannot obtain that the distribution of number of events in interval $[0, t]$ is a new generalized Poisson with parameters $(\theta t, \lambda t)$. Moreover, the definition of new generalized Poisson process is not well-defined.

Suppose the above counting process is a strongly P-process and the distribution of number of events in interval $[0, t]$ is a new generalized Poisson with parameters $(\theta t, \lambda t)$.

Then, by Proposition 1,

$$g_0(t) = -\frac{P_0'(t)}{P_0(t)}$$

and

$$g_n(t) = \frac{g_{n-1}(t)P_{n-1}(t)}{P_n(t)} - \frac{P_n'(t)}{P_n(t)}, \quad n \geq 1.$$

But, since $g_0(t) = \theta$, $P_0(t) = e^{-\theta t}$ and $P_1(t) = \theta t e^{-\theta t - \lambda t}$,

$$g_1(t) = \frac{1}{t}(e^{\lambda t} - 1) + (\theta + \lambda).$$

Thus $g_1(t) = \frac{1}{t}(e^{\lambda t} - 1) + (\theta + \lambda)$ is not equal to the result of Axiom 4.

Hence, the counting process of which the distribution of number of events in interval $[0, t]$ is a new generalized Poisson with parameters $(\theta t, \lambda t)$ is not a P-process. It is because $g_1(t)$ does not satisfy

$$-\infty < \left[\int g_1(t) dt \right]_{t=0} < \infty.$$

Now, we show that Axiom 4 is contradicted to Axiom 3.

From Axiom 3,

$$P\{N(t+h) - N(t) \geq 2 \mid N(t) = x\} = o(h) \quad \text{for each } x = 0, 1, 2, \dots.$$

Note that

$$\begin{aligned} 1 &= \sum_{i=0}^{\infty} P\{N(t+h) - N(t) = i \mid N(t) = x\} \\ &= P\{N(t+h) - N(t) = 0 \mid N(t) = x\} + P\{N(t+h) - N(t) = 1 \mid N(t) = x\} \\ &\quad + P\{N(t+h) - N(t) \geq 2 \mid N(t) = x\}. \end{aligned}$$

Then

$$\begin{aligned} o(h) &= P\{N(t+h) - N(t) \geq 2 \mid N(t) = x\} \\ &= 1 - P\{N(t+h) - N(t) = 0 \mid N(t) = x\} - P\{N(t+h) - N(t) = 1 \mid N(t) = x\}. \end{aligned}$$

Taking $x = 0$, by Axiom 4,

$$\begin{aligned} &P\{N(t+h) - N(t) \geq 2 \mid N(t) = 0\} \\ &= 1 - P\{N(t+h) - N(t) = 0 \mid N(t) = 0\} - P\{N(t+h) - N(t) = 1 \mid N(t) = 0\} \\ &= \theta h - \theta h e^{-\lambda t} + o(h) \\ &\neq o(h). \end{aligned}$$

This is a contradiction to Axiom 3. Therefore, the definition of a generalized Poisson process is not well-defined.

3. Transformed Geometric Poisson process

Let X be a geometric random variable. Then random variable $Y = X - 1$ is called to be *transformed geometric*. In this section, we introduce a P-process for which the distribution of number of events in interval $[0, t]$ is transformed geometric and $g_i(t) \neq g_j(t)$ for each $i \neq j (i, j = 0, 1, 2, \dots)$.

[Definition 3] The P-process $\{N(t) \mid t \geq 0\}$ is said to be a *transformed geometric Poisson process* with intensity function $f(t)$ if

- (i) $f(0) = 0$
- (ii) $0 \leq f(t) < 1$ for each $t \geq 0$
- (iii) $g_n(t) = (n + 1) \frac{df(t)/dt}{1 - f(t)}$.

[Theorem 3] Let $\{N(t) \mid t \geq 0\}$ be a transformed geometric Poisson process with intensity function $f(t)$. Then the number of events in interval $[0, t]$ is a transformed geometric distribution with parameter $(1 - f(t))$ and the integral constants of $\{N(t) \mid t \geq 0\}$ are $k_n = (-1)^n$.

(Proof) Since $P_0(t) = k_0 \exp\left(-\int_* g_0(t) dt\right) = k_0(1 - f(t))$ and

$$g_0(t) = \frac{df(t)/dt}{1 - f(t)}$$

is a t-zero function, $k_0 = 1$ by [Theorem 2].

Hence,

$$P_0(t) = (1 - f(t)).$$

$$\begin{aligned} P_1(t) &= \exp\left(-\int_* g_1(t) dt\right) \left[\int_* g_0(t) P_0(t) \exp\left(\int_* g_1(t) dt\right) + k_1 \exp\left(-\int_* g_1(t) dt\right) \right] \\ &= (1 - f(t))^2 \int_* \frac{f(t)}{(1 - f(t))^2} dt + k_1(1 - f(t))^2 \\ &= (1 - f(t)) + k_1(1 - f(t))^2. \end{aligned}$$

The boundary condition $P_1(0) = 0$ implies that $k_1 = -1$. Then,

$$P_1(t) = f(t)(1 - f(t)).$$

To show that $P_n(t) = f(t)^n(1 - f(t))$, we use the mathematical induction.

Suppose $P_{n-1}(t) = f(t)^{n-1}(1 - f(t))$. Then

$$\begin{aligned} P_n(t) &= \exp\left(-\int_0^t g_n(s) ds\right) \left[\int_0^t g_{n-1}(s) P_{n-1}(s) \exp\left(\int_0^s g_n(u) du\right) ds \right] + k_n \exp\left(-\int_0^t g_n(s) ds\right) \\ &= (1 - f(t))^{n+1} \int_0^t \frac{nf(s)f(s)^{n-1}}{(1 - f(s))^{n+1}} ds + k_n(1 - f(t))^{n+1} \\ &= \sum_{i=1}^n (-1)^{i+1} f(t)^{n-i}(1 - f(t))^i + k_n(1 - f(t))^{n+1}. \end{aligned}$$

The boundary condition $P_n(0) = 0$ implies that $k_n = (-1)^n$. Hence, we obtain

$$P_n(t) = f(t)^n(1 - f(t)). \quad \square$$

[Example 3] Let $f(t) = 1 - e^{-\theta t}$. Then $f(0) = 0$ and $0 \leq f(t) < 1$.

Taking $g_n(t) = (n + 1) \frac{df(t)/dt}{1 - f(t)}$,

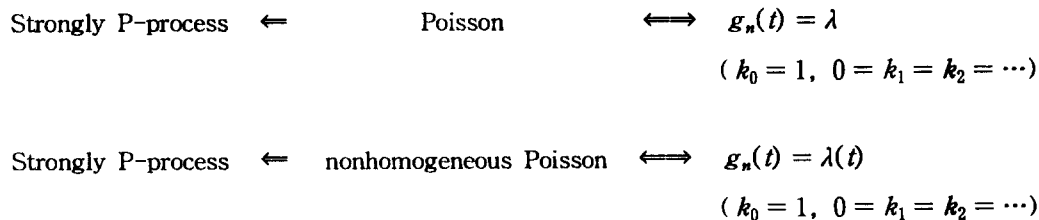
$$g_n(t) = (n + 1)\theta$$

which implies

$$P_n(t) = (1 - e^{-\theta t})^n e^{-\theta t}.$$

Therefore, the number of events in interval $[0, t]$ is a transformed geometric distribution with parameter $f(t) = 1 - e^{-\theta t}$. \square

The following diagram represents that the relation of the generalized Poisson process with $g_n(t)$ and the integral constants.



$$\text{Strongly P-process} \leftarrow (0,1)\text{-generalized Poisson} \iff g_n(t) = \begin{cases} \theta - \frac{\theta\lambda}{1+\theta\lambda t}, & n=0 \\ \frac{\theta}{1-\lambda}, & n=1 \\ \theta, & n \geq 2 \end{cases}$$

$$(k_0 = 1, 0 = k_1 = k_2 = \dots)$$

$$\text{P-process} \leftarrow (1, 2)\text{-generalized Poisson} \iff g_n(t) = \begin{cases} \theta - \frac{2\theta\lambda}{2+\theta\lambda t}, & n=1 \\ \frac{\theta}{1-\lambda}, & n=2 \\ \theta, & n=0, 3, 4, \dots \end{cases}$$

$$(k_0 = 1, k_1 = 1/2\lambda, 0 = k_2 = \dots)$$

$$\text{P-process} \leftarrow \text{transformed geometric Poisson} \iff g_n(t) = (n+1) \frac{df(t)/dt}{1-f(t)}$$

$$(k_n = (-1)^n).$$

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