

93 Conditional or Mixed PP

Draw a λ at random from a dist'n.

Then form the PP (λ).

In other words, there is a positive r.v. L such that, conditional on $L = \lambda$,

$N(t)$ is a PP with parameter λ .

This process is called conditional or mixed Poisson process. (L is the mixing dist'n.)

$$P(N(t+s) - N(s) = m) =$$

$$\int P(N(t+s) - N(s) = m \mid L = \lambda) g(\lambda) d\lambda = (*)$$

where g is the (continuous) density

of L .

If L takes two values: λ_1 and λ_2 with prob. p and $q = 1-p$ we get

$$P(N(t+s) - N(s) = m) =$$

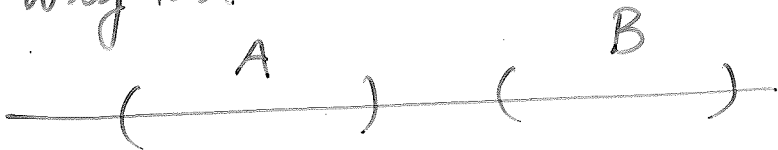
$$p \cdot e^{-\lambda_1 t} \frac{(\lambda_1 t)^m}{m!} + (1-p) e^{-\lambda_2 t} \frac{(\lambda_2 t)^m}{m!}$$

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$$(*) = \int e^{-\lambda t} \frac{(\lambda t)^n}{n!} \cdot g(\lambda) d\lambda \quad (5.27)$$

A mixed PP is not a PP in general.*
It has stationary increments, though, as shown by (5.27).

* Why not?



The no. of points in A gives no indications as to what the value of λ is, affecting the probability, of the no. of points in B.
dist'n

Ex. Assume we have two kinds of customers in an insurance company. Type I are accident-prone $\lambda_1 = 10$, Type II are prudent, $\lambda_2 = 1$. Probability of a random customer being of type I or II is equal, $= \frac{1}{2}$.

Assume $N(1) = 10$, calculate $E(N(2) - N(1))$.

$N(1) = 10$ given.

$$P(\lambda = \lambda_1 | N(1) = 10) = \frac{P(\lambda = \lambda_1 \text{ and } N(1) = 10)}{P(N(1) = 10)} =$$

$$= \frac{e^{-10} \frac{10^{10}}{10!}}{P(N(1) = 10, \lambda = \lambda_1) + P(N(1) = 10, \lambda = \lambda_2)}$$

$$= \frac{e^{-10} \cdot \frac{10^{10}}{10!}}{e^{-10} \cdot \frac{10^{10}}{10!} + e^{-1} \cdot \frac{1}{10!}} \equiv p_1$$

$$p_2 \equiv 1 - p_1$$

$$\mathbb{E}(N(2) - N(1) \mid \lambda = \lambda_1) = 10 \quad \text{and}$$

$$\mathbb{E}(N(2) - N(1) \mid \lambda = \lambda_2) = 1$$

$$\mathbb{E}(N(2) - N(1) \mid N(1) = 10) = p_1 \cdot 10 + p_2 \cdot 1$$

Example 5.29

$$g(\lambda) = \theta e^{-\theta \lambda} \frac{(\theta \lambda)^{m-1}}{(m-1)!} \quad \lambda > 0$$

gamma (m, θ)

In this case $N(t)$ has a dist'n which can be explicitly calculated:

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$$\int_0^{\infty} e^{-\lambda t} \frac{(\lambda t)^n}{n!} \theta e^{-\theta \lambda} \frac{(\theta \lambda)^{m-1}}{(m-1)!} d\lambda$$

$$= \frac{1}{n!(m-1)!} t^n \theta^m \int_0^{\infty} e^{-(t+\theta)\lambda} \lambda^{n+m-1} d\lambda$$

$$= \frac{1}{n!(m-1)!} \frac{t^n \theta^m (n+m-1)!}{(t+\theta)^{n+m}}$$

$$\int_0^{\infty} \frac{(t+\theta) e^{-(t+\theta)\lambda} \lambda^{n+m-1} (t+\theta)^{n+m-1} d\lambda}{(n+m-1)!}$$

↑ density of Gamma ($n+m$, $t+\theta$)

$$= \binom{n+m-1}{n} \left(\frac{\theta}{t+\theta}\right)^m \left(\frac{t}{t+\theta}\right)^n \cdot 1$$

negative binomial distr.

with param. $\frac{\theta}{t+\theta}$

= no. of failures before n successes
in a binomial trial with

$p = \frac{\theta}{t+\theta}$ prob. of success.

97 Influence of observations: If we observe $N(t)$ for some time we get better and better estimate of the λ chosen.

$f_{L|N(t)}$ the cond. density of L given $N(t)$
sought

$$E(N(t)|L) = Lt \quad \text{gives } EN(t) = E(L) \cdot t$$

$$\text{Var}(N(t)|L) = Lt$$

$$\text{Var}(N(t)) = E(L)t + t^2 \text{Var}(L)$$

Why?

$$\text{Var}(N(t)) = E(N(t)^2) - \underbrace{(E(N(t)))^2}_{t^2(E(L))^2}$$

$$\begin{aligned} E(N(t)^2) &= E(E(N(t)^2|L)) \\ &= E(Lt + L^2 t^2) = t E(L) + t^2 E(L^2) \\ &= t E(L) + t^2 (\text{Var}(L) + (E(L))^2) \end{aligned}$$

so

$$\text{Var}(N(t)) = t E(L) + t^2 \text{Var}(L)$$

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 $f_{L|N(t)} ?$

$$P(L \leq x | N(t) = m) = \frac{P(L \leq x, N(t) = m)}{P(N(t) = m)}$$

$$= \frac{\int_0^{\infty} P(L \leq x, N(t) = m | L = \lambda) g(\lambda) d\lambda}{P(N(t) = m)}$$

$$= \frac{\int_0^x P(N(t) = m | L = \lambda) g(\lambda) d\lambda}{P(N(t) = m)}$$

$$= \frac{\int_0^x e^{-\lambda t} \frac{(\lambda t)^m}{m!} g(\lambda) d\lambda}{\int_0^{\infty} e^{-\lambda t} \frac{(\lambda t)^m}{m!} g(\lambda) d\lambda}$$

density! differentiate w.r.t. x

$$e^{-xt} \cdot (xt)^m g(x)$$

$$\int_0^{\infty} e^{-\lambda t} \frac{(\lambda t)^m}{m!} g(\lambda) d\lambda$$

99 Evaluate at $x=1$

$$f_{L|N}(t) (\lambda | n) =$$

$$(5.28) \quad \frac{e^{-\lambda t} \lambda^n g(\lambda)}{\int_0^{\infty} e^{-\lambda t} \lambda^n g(\lambda) d\lambda} \quad \lambda \geq 0$$

Discrete case $P(L = \lambda_j) = g_j \quad j=1, 2, \dots$

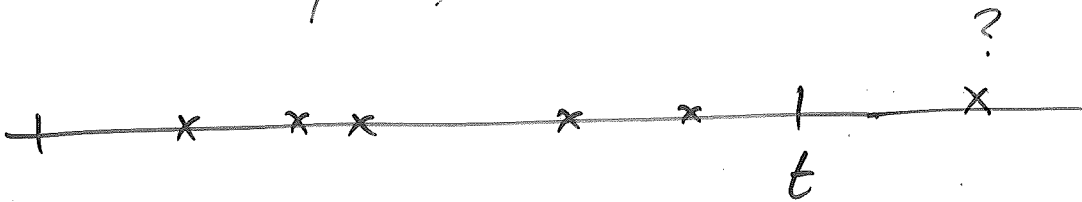
$$\frac{e^{-\lambda_j t} \lambda_j^n g_j}{\sum_{j=1}^{\infty} e^{-\lambda_j t} \lambda_j^n g_j}$$

Case $\lambda_1 = \lambda_2 = \frac{1}{2}$

$$\frac{e^{-\lambda_j t} \lambda_j^n}{e^{-\lambda_1 t} \lambda_1^n + e^{-\lambda_2 t} \lambda_2^n}$$

Ex. 5.30 Insurance company. Customer rating " λ " means that claims will be initiated at the rate λ (per year). The prob. dist'n of λ is uniform over $(0, 1)$. Given that a customer has made n claims during his/her first t years, what is the cond'l dist'n of the time to the next claim.

Note. If we know the rate is λ then the time is $\text{Exp}(\lambda)$.



$$P(T > x \mid N(t) = n)$$

T is the time until the next claim

$$= \int_0^1 P(T > x \mid L = \lambda, N(t) = n) f_{L \mid N(t)}(\lambda \mid n)$$

$$= \frac{\int_0^1 e^{-\lambda x} \cdot e^{-\lambda t} \cdot \lambda^n d\lambda}{\int_0^1 e^{-\lambda t} \lambda^n d\lambda}$$

$$g(\lambda) = 1 \\ 0 < \lambda < 1$$