

# Probability and Stochastic Processes

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## Problem 4:

With  $P_n = P(X = n)$ , let  $G_E(x) = p_0 + p_2(x^2) + p_4(x^4) + \dots$  and

$G_O = p_1x + p_3(x^3) + p_5(x^5) + \dots$  be the even and odd components of the probability generating function  $G(x)$ .

One of the given conditions was  $p_{2n} = \frac{2}{3}p_{2n+1}$  which gives,

$$G_E(x) = \frac{2}{3}p_1 + \frac{2}{3}p_3x^2 + \frac{2}{3}p_5x^4 + \dots = \frac{2}{3}G_O(x)/x \quad (1)$$

On the other hand,  $p_{2n} = \frac{1}{2}p_{2n-1}$  for all  $n$  greater than 1. It gives us that,

$$G_E(x) = p_0 + \frac{1}{2}p_1(x^2) + \frac{1}{2}p_3x^4 + \dots = p + \frac{1}{2}xG_O(x) \quad (2)$$

On combining 1 and 2, we get,

$$p_0x + \frac{1}{2}x^2G_O(x) = \frac{2}{3}G_O(x) \quad (3)$$

$$\implies G_O(x) = \frac{p_0x}{\frac{2}{3} - \frac{1}{2}x^2} = \frac{6p_0x}{4 - 3x^2} \quad (4)$$

and therefore  $G_E(x) = \frac{4p_0}{4 - 3x^2}$ , so that

$$G(x) = G_E(x) + G_O(x) = \frac{p_0(6x + 4)}{4 - 3x^2} = \frac{1}{5} \frac{3x + 2}{4 - 3x^2} \quad (5)$$

Because any PGF must satisfy  $G(x) = 1$ , we get,  $p_0 = \frac{1}{10}$

## Problem 5:

(A. Gut, first ed., III.15) The number of cars passing a road crossing during a day follows a Poisson distribution with parameter  $\lambda$ . The number of persons in each car is a  $\text{Poisson}(\alpha)$  random variable. Find the probability generating function of the total number of persons,  $N$ , passing the road crossing during a day. Find the mean and variance of  $N$ .

Consider that, If  $X$  is the number of cars and  $Y$  the number of persons in each car, then the number of total persons is equal to the product of the total number of cars ( $X$ ) and the total number of persons in each car ( $Y$ ). So, if we denote the total number of persons by  $N$ , then we will have the following-

$$G_N(s) = G_X(G_Y(s))$$

where X and Y are the total number of cars and total number of persons in each of the cars. Now,

$$G_N(s) = G_X(G_Y(s)) \implies e^{\lambda((e^{\alpha(s-1)})-1)}$$

Now, our objective is to find the expected number of persons crossing the road on a given day. So, we define the expectation of a random variable X as follows-

$$E(X) = G'(1)$$

and,

$$G'(N) = e^{\lambda((e^{\alpha(s-1)})-1)} \lambda \frac{d(e^{\alpha(s-1)})}{ds}$$

or,

$$G'(N) = e^{\lambda((e^{\alpha(s-1)})-1)} \lambda (e^{\alpha(s-1)}) \alpha$$

Similarly,

$$G''(N) = \lambda \alpha \{ e^{\lambda((e^{\alpha(s-1)})-1)} \lambda (e^{\alpha(s-1)}) \alpha + (e^{\alpha(s-1)}) \} \alpha$$

or,

$$G''(N) = (\lambda \alpha)^2 (e^{\lambda((e^{\alpha(s-1)})-1)}) (e^{\alpha(s-1)}) + \lambda \alpha^2 (e^{\alpha(s-1)})$$

So, now when we have the  $G'(N)$ , we can proceed towards finding the value of  $G'(1)$ , that is,

$$G'(1) = \lambda \alpha$$

So, we can say that the mean number of person crossing the road is equal to the  $\lambda \alpha$ . Now, it would be a great idea to estimate the variance which is define as follows-

$$Var(N) = E[(N - E(N))^2] = E(N^2) - (E(N))^2$$

But,

$$E(N(N-1)) = G''(1)$$

or,

$$E(N^2) = G''(1) + E(N)$$

or,

$$E(N^2) = \lambda \alpha^2 (\lambda + 1) + \lambda \alpha$$

or,

$$E(N^2) = \lambda \alpha (\lambda \alpha + \alpha + 1)$$

so,

$$Var(N) = E(N^2) - (E(N))^2 = (\lambda\alpha)^2 + \lambda\alpha^2 + \lambda\alpha - (\lambda\alpha)^2$$

or,

$$Var(N) = \lambda\alpha(\alpha + 1)$$

**Problem 10:** Using a moment generating function to prove that if a random variable  $X$  has density function

$$f_X(x) = \frac{1}{2}e^{-|x|}, \quad -\infty < x < \infty, \quad (6)$$

Bf then  $X$  can be written as  $X = YZ$  where  $Y$  and  $Z$  are independent random variables exponentially distributed.

We know that if  $Y \sim Exp(\mu)$ , then:

$$\Phi_Y(t) = \int_{-\infty}^{\infty} e^{ty} f_Y(y) dy = \int_0^{\infty} \mu e^{-(\mu-t)y} dy = \frac{\mu}{\mu-t}, \quad t < \mu$$

Moreover, if  $Z \sim Exp(\lambda)$ , consider the moment generating function of  $-Z$ :

$$\begin{aligned} \Phi_{-Z}(t) &= \int_{-\infty}^{\infty} e^{-tz} f_Z(z) dz = \int_0^{\infty} e^{-tz} \lambda e^{-\lambda z} dz = \int_0^{\infty} \lambda e^{-z(t+\lambda)} dz = \int_0^{\infty} \lambda e^{-z(t+\lambda)} dz \\ &= \left[ \frac{-\lambda}{t+\lambda} e^{-z(t+\lambda)} \right]_0^{\infty} \end{aligned}$$

and if  $t > -\lambda$ :

$$= \frac{\lambda}{t+\lambda}$$

By the convolution theorem (seen in class), we know that if  $S = X_1 + \dots + X_n$ , then:

$$\Phi_S(t) = \prod_{k=1}^n \phi_{X_k}(t)$$

Then, in our case, we show that if  $-\lambda < t < \mu$ :

$$\Phi_X(t) = \phi_Y Z(t) = \phi_Y(t) \phi_{-Z}(t) = \frac{\mu}{\mu-t} \frac{\lambda}{t+\lambda}$$

On the other hand, let's see what the generating function of moments of our variable  $X$  is:

$$\begin{aligned} \Phi_X(t) &= \int_{-\infty}^{\infty} e^{tx} f_X(x) dx = \int_{-\infty}^{\infty} e^{tx} \frac{1}{2} e^{-|x|} dx = \frac{1}{2} \left( \int_{-\infty}^0 e^{tx} e^x dx + \int_0^{\infty} e^{tx} e^{-x} dx \right) \\ &= \frac{1}{2} \left( \int_{-\infty}^0 e^{x(t+1)} dx + \int_0^{\infty} e^{x(t-1)} dx \right) = \frac{1}{2} \left( \left[ \frac{1}{t+1} e^{x(t+1)} \right]_{-\infty}^0 + \left[ \frac{1}{t-1} e^{x(t-1)} \right]_0^{\infty} \right) = \end{aligned}$$

$$= \frac{1}{2} \left( \frac{1}{t+1} - \lim_{x \rightarrow -\infty} \frac{1}{t+1} e^{x(t+1)} \right)_{(1)} + \frac{1}{2} \left( \lim_{x \rightarrow \infty} \frac{1}{t-1} e^{x(t-1)} - \frac{1}{t-1} \right)_{(2)}$$

Evaluated separately (1) and (2). If  $t > -1$ :

$$(1) = \frac{1}{2} \left( \frac{1}{t+1} - \lim_{x \rightarrow -\infty} \frac{1}{t+1} e^{x(t+1)} \right) = \frac{1}{2(t+1)}$$

Moreover, if  $t < 1$ :

$$(2) = \frac{1}{2} \left( \lim_{x \rightarrow \infty} \frac{1}{t-1} e^{x(t-1)} - \frac{1}{t-1} \right) = \frac{-1}{2(t-1)} = \frac{1}{2(1-t)}$$

. Thus if  $-1 < t < 1$ , then:

$$\Phi_X(t) = \frac{1}{2} \left( \frac{1}{t+1} + \frac{1}{1-t} \right) = \frac{1}{2} \left( \frac{1-t+t+1}{(t+1)(1-t)} \right) = \frac{1}{2} \left( \frac{2}{(t+1)(1-t)} \right) = \frac{1}{(t+1)(1-t)}$$

If we recall the expression that we said before we wanted to get there:

$$\Phi_X(t) = \frac{\mu}{\mu-t} \frac{\lambda}{t+\lambda}, \quad -\lambda < t < \mu$$

Then we are considering  $\lambda = 1$  and  $\mu = 1$ :

$$\Phi_X(t) = \frac{\mu}{\mu-t} \frac{\lambda}{t+\lambda} = \frac{1}{1-t} \frac{1}{t+1}$$

And so we have shown that  $X$  can be written as  $X = YZ$ , where  $Y$  and  $Z$  are exponential.

**Problem 4: 5. (A. Gut's book)** The following model can be used to describe the number of women (mothers and daughters) in a given area. The number of mothers is a random variable  $X \sim \text{Poisson}(\lambda)$ . Independently of the others, every mother gives birth to a  $\text{Poisson}(\mu)$ -distributed number of daughters. Let  $Y$  be the total number of daughters and hence  $Z = X + Y$  be the total number of women in the area.

Let  $Y_k$  be the count of daughters for mother  $k$  so that

$$Y = \sum_{k=1}^X Y_k \tag{7}$$

The generating function of a sum of random variables is equal to the product of their generating functions only when the random variables are independent  $Y = \sum_{k=1}^X Y_k$  means that  $Y$  is rather dependent on  $X$ , so the step,  $G_{(X+Y)}(s) = G_X(s) + G_Y(s)$

Instead, return to the definition of probability generation, and use the identical distribution of all  $Y_k$  and the total independence of all  $Y_k$  and  $X$ .

$$G_Z(s) = E(s^Z) = E(E(s^{X+\sum_{k=1}^X Y_k} | X)) \tag{8}$$

$$\implies G_Z(s) = G_X(sG_{Y_1}(s)) = \exp(\lambda(se^{\mu(s-1)} - 1)) \quad (9)$$

Differentiating it with respect to  $Z$  and substituting 1, we get,  $G'_Z(s) = \exp(\lambda(se^{\mu(s-1)} - 1)) \cdot \lambda (s \cdot e^{\mu(s-1)}) \mu + 1$

Setting,  $s=1$ , we get,

$$E(Z) = G'(1) = \lambda(\mu + 1)$$

Similarly, we differentiate  $G'(Z)$  to get  $G''(Z)$  and set  $s=1$  to get  $\text{Var}(Z) = \lambda(1 + 3\mu + \mu^2)$

**Problem 5:**