

## Stochastic Processes (1)

### Lecture 6: P.G.F. of Sum of Random Number of r.v.'s

#### (1-5) P.G.F. of Sum of Random Number of Discrete r.v.'s:

To explain this subject we take this example, if  $X_i$  denotes to the number of person's involved the ( $i^{th}$ ) accident in a day in a certain city, and if the number ( $N$ ) of accident happening on a day is a random variable, then the sum:

$S_n = \sum_{i=1}^n X_i$  denoted the total number of persons involved in accidents on a day.

#### **Theorem (3):**

Let  $X_i$ ,  $i = 1, 2, \dots, n$  be (i.i.d.) random variables with  $P_r\{X_i = k\} = p_k$ , and p.g.f. of  $X_i$  is:

$$P_{X_i}(S) = \sum_{k=0}^{\infty} p_k S^k$$

and let the sum:

$S_N = X_1 + X_2 + \dots + X_N$ , where  $N$  is a random variable independent of the  $X_i$ 's. Let the distribution of ( $N$ ) is given by:

$P_r\{N = n\} = g_n$ , where the p.g.f. of  $N$  is given by:

$$G_N(S) = \sum_{n=0}^{\infty} g_n S^n \quad \dots 1.14$$

Then  $H(S)$  is the p.g.f. of the sum  $S_N$ , and it given by compound function:

$$H(S) = \sum_{j=0}^{\infty} h_j S^j = \sum_{j=0}^{\infty} P_r\{S_N = j\} S^j, \text{ Where } h_j = P_r\{S_N = j\}$$

$$\text{Then: } H(S) = G_N[P_X(S)] \quad \dots 1.15$$

**Proof:**

Since  $N$  is a random variable which can assume values  $(0, 1, 2, \dots)$ , then the event:  $\{S_N = j\}$  can be happen in the mutually exclusive ways:

$$[N = n \text{ and } S_n = X_1 + X_2 + \dots + X_n = j], \text{ for } n = 1, 2, \dots$$

Define that  $X_0 = S_0 = 0$ , so that:

$$X_0 + X_1 + X_2 + \dots + X_n = S_n, \quad n \geq 0$$

Then:

$$h_j = P_r\{S_N = j\} = \sum_{n=0}^{\infty} P_r\{N = n, S_n = j\}$$

Since  $N$  is independent of  $X_i$ 's and therefor of  $S_n$ , then:

$$\begin{aligned} h_j &= \sum_{n=0}^{\infty} P_r\{N = n\} P_r\{S_n = j\} \\ &= \sum_{n=0}^{\infty} g_n P_r\{S_n = j\} \end{aligned}$$

Then the p.g.f. of the sum  $S_N = X_1 + X_2 + \cdots + X_N$  is given by:

$$\begin{aligned} H(S) &= \sum_{j=0}^{\infty} h_j S^j = \sum_{j=0}^{\infty} \left[ \sum_{n=0}^{\infty} g_n P_r\{S_n = j\} S^j \right] \\ &= \sum_{n=0}^{\infty} \left[ \sum_{j=0}^{\infty} P_r\{S_n = j\} S^j \right] g_n \end{aligned}$$

Since:  $S_n = X_1 + X_2 + \cdots + X_n$  is (i.i.d) r.v's of sum of fixed number  $n$  [from theorem (2)], then:

$$\begin{aligned} H(S) &= \sum_{n=0}^{\infty} [P_X(S)]^n g_n \\ &= G_N[P_X(S)] \quad , \text{ compound function.} \end{aligned}$$

Note: if  $P_X(S)$  and  $G_N(S)$  are two p.g.f. , then the  $G_N[P_X(S)]$  is also p.g.f.

**Corollary (1):**

The mean and variance of the sum  $S_N = \sum_{i=1}^N X_i$ , (where  $N$  is a random variable) are :

$$1. E(S_N) = E(N).E(X_i) \quad \dots 1.16$$

$$2. var(S_N) = E(N).var(X_i) + [E(X_i)]^2.var(N) \quad \dots 1.17$$

**Proof:**

1. Since  $H(S)$  is the p.g.f. of the sum  $S_N$ , then:

$$H(S) = G_N(P_X(S)) \quad , \quad [\text{theorem (3)}]$$

Then the mean of  $S_N$  is:

$$\begin{aligned} E(S_N) &= H'(1) \\ &= G'_N(P_X(S)).P'_X(S)|_{S=1} \\ &= G'_N(1).P'_X(1) \quad , \quad \text{since } P_X(1) = 1 \\ &= E(N).E(X_i) \quad . \end{aligned}$$

2. Since  $H(S)$  is the p.g.f. of the sum  $S_N$ , then:

$$var(S_N) = H''(1) + H'(1) - [H'(1)]^2 \quad , \quad (\text{from the definition of variance})$$

Since:

$$\begin{aligned} H''(1) &= P'_X(S).G''_N(P_X(S)).P'_X(S)|_{S=1} + G'_N(P_X(S)).P''_X(S)|_{S=1} \\ &= [P'_X(1)]^2.G''_N(1) + G'_N(1).P''_X(1) \end{aligned}$$

Then:

$$\begin{aligned}
 \text{var}(S_N) &= [P'_X(1)]^2 \cdot G''_N(1) + G'_N(1) \cdot P''_X(1) + G'_N(1) \cdot P'_X(1) \\
 &\quad - [G'_N(1) \cdot P'_X(1)]^2 \\
 &= G'_N(1)[P''_X(1) + P'_X(1)] + [P'_X(1)]^2[G''_N(1) - [G'_N(1)]^2] \\
 &\quad \dots 1.18
 \end{aligned}$$

By adding and subtracting the amount  $[P'_X(1)]^2 G'_N(1)$  to the right side of the equation (1.18), we have:

$$\begin{aligned}
 \text{var}(S_N) &= G'_N(1)[P''_X(1) + P'_X(1)] - [P'_X(1)]^2 G'_N(1) \\
 &\quad + [P'_X(1)]^2 G'_N(1) + [P'_X(1)]^2[G''_N(1) - [G'_N(1)]^2] \\
 \text{var}(S_N) &= G'_N(1)[P''_X(1) + P'_X(1) - [P'_X(1)]^2] \\
 &\quad + [P'_X(1)]^2[G''_N(1) + G'_N(1) - [G'_N(1)]^2]
 \end{aligned}$$

$$\text{var}(S_N) = E(N) \cdot \text{var}(X_i) + [E(X_i)]^2 \cdot \text{var}(N),$$

Proof of theorem.

**Example (1.13):**

If  $S_N = X_1 + X_2 + \dots + X_N$  where  $N$  has a Poisson distribution with mean  $\lambda$ , and if  $X_i$  are (i.i.d) random variables with Geometric distribution:

$P_r\{X_i = k\} = pq^k$ , find the p.g.f. of the sum  $S_N$  and the mean and variance of it.

**Solution:**

Since:  $N \sim \text{poi}(\lambda)$ , then the p.g.f. of  $N$  is:

$$G_N(S) = e^{\lambda(S-1)}, \text{ with mean and variance } \lambda.$$

And since:  $X_i \sim \text{Geo}(p)$ , then the p.g.f. of  $X_i$  is:

$$P_{X_i}(S) = \frac{p}{1-qS}, \text{ with mean } \left(\frac{q}{p}\right) \text{ and variance } \left(\frac{q}{p^2}\right).$$

Then the p.g.f. of  $S_N = X_1 + X_2 + \dots + X_N$  is:

$$H(S) = G_N[P_{X_i}(S)], \quad [\text{theorem (3)}]$$

$$= G_N\left[\frac{p}{1-qS}\right]$$

$$= e^{\lambda\left(\frac{p}{1-qS}-1\right)}, \text{ the p.g.f. of } S_N.$$

The mean of  $S_N$  is:

$$E(S_N) = E(N) \cdot E(X_i) = \lambda \frac{q}{p} = \frac{\lambda q}{p}$$

The variance of  $S_N$  is:

$$\text{var}(S_N) = E(N) \cdot \text{var}(X_i) + [E(X_i)]^2 \cdot \text{var}(N)$$

$$= \lambda \frac{q}{p^2} + \left(\frac{q}{p}\right)^2 \lambda$$

$$= \frac{\lambda q}{p^2} (1 + q).$$

**Example (1.14):**

Let  $X_1 + X_2 + \dots + X_N$  be (i.i.d) random variables with Poisson distribution ( $\lambda$ ), find the p.g.f. of:

$S_N = \sum_{i=1}^N X_i$  , where:

1.  $n$  is fixed integer.
2.  $N$  is random variable and has a Binomial distribution with parameter  $(n, p)$ .

**Solution:**

1. Since  $X_i$  is Poisson distribution with the same parameter ( $\lambda$ ), then the p.g.f. of  $X_i$  is:

$$P_{X_i}(S) = e^{\lambda(S-1)}, \text{ for all } X_i.$$

Then the p.g.f. of sum i.i.d. ( $S_N$ ) is:

$$\begin{aligned} P_{S_n}(S) &= [P_X(S)]^n, \quad [\text{theorem (2)}] \\ &= [e^{\lambda(S-1)}]^n = e^{n\lambda(S-1)} \end{aligned}$$

2. Since  $N$  is Binomial distribution with parameters  $(n, p)$ ,

then the p.g.f. of the random variable  $N$  is:

$$G_N(S) = \sum_{n=0}^{\infty} g_n S^n = (pS + q)^n$$

Then the p.g.f. of  $(S_N)$  when  $N$  is r.v., is:

$$H(S) = G_N[P_{X_i}(S)] \quad , \quad [\text{Theorem (3)}]$$

$$= G_N[e^{\lambda(S-1)}]$$

$$= (pe^{\lambda(S-1)} + q)^n \quad , \quad \text{the p.g.f. of } S_N.$$

*Remark:* To find the mean and variance of  $S_N$  when  $N$  is random variable:

$$E(S_N) = H'(S)$$

$$= E(N) \cdot E(X_i) = np\lambda \quad , \quad \text{the mean of } S_N.$$

The variance of  $S_N$  is:

$$\text{var}(S_N) = E(N) \cdot \text{var}(X_i) + [E(X_i)]^2 \cdot \text{var}(N)$$

$$= np\lambda + \lambda^2 npq = \lambda np(1 + \lambda q) \quad .$$

**H.W:** From example (1.14), find the mean and variance of  $S_N$

when:  $X_i \sim \text{Poi}(2)$  and  $N \sim \text{Bin}(30, \frac{1}{2})$  .