

## Tutorial Session ④

### Random Sums

#### Theorem

For the random sum  $X = \mathcal{J}_1 + \mathcal{J}_2 + \dots + \mathcal{J}_k + \dots + \mathcal{J}_N$ ,  $N > 0$

$$\text{If } E(\mathcal{J}_k) = \mu, \quad \text{Var}(\mathcal{J}_k) = \sigma^2$$

$$E(N) = \nu, \quad \text{Var}(N) = \tau^2$$

then

$$E(X) = \mu\nu \quad \text{and} \quad \text{Var}(X) = \nu\sigma^2 + \mu^2\tau^2$$

#### \* pb 2.3.5 p.64 Textbook

Let  $N$  is # accidents in a week, and  $\mathcal{J}_k$  is # individuals injured for  $k^{\text{th}}$  accident.  $X = \mathcal{J}_1 + \mathcal{J}_2 + \dots + \mathcal{J}_N$ ,  $N > 0$

$$N \sim \text{poisson}(2)$$

$$E(N) = 2, \quad \text{Var}(N) = 2$$

$$\mathcal{J}_k: E(\mathcal{J}_k) = 3, \quad \text{Var}(\mathcal{J}_k) = 4$$

$$\therefore E(X) = \mu\nu = 3(2) = 6,$$

$$\text{Var}(X) = \nu\sigma^2 + \mu^2\tau^2 = 2(4) + 9(2) = 26 \quad \#$$

#### \* pb 2.3.5 p.65 Textbook

$$\therefore E(\mathcal{J}_k) = \mu, \quad \text{Var}(\mathcal{J}_k) = \sigma^2$$

$$\text{and } \therefore E(N+1) = E(N) + 1 = \nu + 1,$$

$$\text{Var}(N+1) = \text{Var}(N) + \text{Var}(1) = \tau^2 + 0 = \tau^2$$

$\therefore$  For the random sum  $X = \mathcal{J}_0 + \mathcal{J}_1 + \dots + \mathcal{J}_N$

$$E(X) = \mu(\nu+1),$$

$$\text{Var}(X) = (\nu+1)\sigma^2 + \mu^2\tau^2$$

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pb. 2.3.3 p. 65

Suppose that  $\xi_1, \xi_2, \dots$  are independent and identically distributed with  $\text{pr}(\xi_k = \pm 1) = \frac{1}{2}$  let  $N$  be independent of  $\xi_1, \xi_2, \dots$  and follow the geometric probability mass function

$$p(k) = \alpha(1-\alpha)^k \text{ for } k=0, 1, \dots$$

where  $0 < \alpha < 1$ . Form the random sum  $Z = \xi_1 + \xi_2 + \dots + \xi_N$

Determine the mean and Variance of  $Z$ .

Ans: we have

$\xi_k$	-1	1
$\text{pr}(\xi_k)$	$\frac{1}{2}$	$\frac{1}{2}$

$$\therefore \mu = E(\xi_k) = \sum \xi_k \text{pr}(\xi_k)$$

$$\therefore \mu = -1\left(\frac{1}{2}\right) + 1\left(\frac{1}{2}\right) = 0$$

$$E(\xi_k^2) = (-1)^2\left(\frac{1}{2}\right) + 1^2\left(\frac{1}{2}\right) = 1$$

$$\text{and } \sigma^2 = \text{Var}(\xi_k) = E(\xi_k^2) - [E(\xi_k)]^2 = 1 - 0 = 1$$

$\therefore N \sim \text{geom}(\alpha)$ ,  $p_N(k) = \alpha(1-\alpha)^k$ ,  $k=0, 1, \dots$

$$\therefore \nu = E(N) = \frac{1-\alpha}{\alpha}, \quad \tau^2 = \text{Var}(N) = \frac{1-\alpha}{\alpha^2}$$

$$\therefore Z = \xi_1 + \xi_2 + \dots + \xi_N$$

$$\therefore E(Z) = \mu\nu = 0\left(\frac{1-\alpha}{\alpha}\right) = 0$$

$$\text{and } \text{Var}(Z) = \nu\sigma^2 + \mu^2\tau^2 = \frac{1-\alpha}{\alpha}(1) + 0\left(\frac{1-\alpha}{\alpha^2}\right)$$

$$\therefore \text{Var}(Z) = \frac{1-\alpha}{\alpha}$$

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$X \sim \text{geom}(p)$   
 $\Rightarrow \text{pr}(X=k) = p(1-p)^k, k=0, 1, \dots$   
 $\Rightarrow E(X) = \frac{1-p}{p}$   
 $\text{Var}(X) = \frac{1-p}{p^2}$

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pb 2.4.3 p. 70

Let  $X$  have a Poisson distribution with parameter  $\lambda > 0$ . Suppose  $\lambda$  itself is random, following an exponential density with parameter  $\theta$ .

- (a) What is the marginal distribution of  $X$ ?  
(b) Determine the conditional density for  $\lambda$  given  $X = k$ .

Ans:

$$\begin{aligned} \text{(a)} \quad \Pr(X=k) &= \int_0^{\infty} \Pr(k|\lambda) f(\lambda) d\lambda \\ &= \int_0^{\infty} \frac{e^{-\lambda} \lambda^k}{k!} \theta e^{-\theta \lambda} d\lambda \\ &= \frac{\theta}{k!} \int_0^{\infty} \lambda^k e^{-(1+\theta)\lambda} d\lambda \end{aligned}$$

$$\begin{aligned} \Pr(X=k) &= \frac{\theta}{k!} \int_0^{\infty} \left(\frac{u}{1+\theta}\right)^k e^{-u} \frac{du}{1+\theta}, \quad u = (1+\theta)\lambda \\ &= \frac{\theta}{k!} \left(\frac{1}{1+\theta}\right)^k \frac{1}{1+\theta} \int_0^{\infty} u^k e^{-u} du \end{aligned}$$

$$\begin{aligned} &= \frac{\theta}{k!} \left(\frac{1}{1+\theta}\right)^k \left(\frac{1}{1+\theta}\right)^{\Gamma(k+1)} \\ &= \frac{\theta}{k!} \left(\frac{1}{1+\theta}\right)^k \left(\frac{1}{1+\theta}\right) k! \end{aligned}$$

$X \sim \text{geom}(p)$   
 $\Rightarrow \Pr(X=k) = p(1-p)^k$   
 $k=0, 1, 2, \dots$

$$\Pr(X=k) = \left(\frac{\theta}{1+\theta}\right) \left(\frac{1}{1+\theta}\right)^k, \quad k=0, 1, \dots$$

which is the prob. mass fn for geometric distn with parameter  $\frac{\theta}{1+\theta}$

i.e.  $X \sim \text{geom}\left(\frac{\theta}{1+\theta}\right)$   $\neq$

$$\begin{cases} p \rightarrow \frac{\theta}{1+\theta} \\ 1-p \rightarrow 1 - \frac{\theta}{1+\theta} \\ \quad = \frac{1+\theta-\theta}{1+\theta} \\ \quad = \frac{1}{1+\theta} \end{cases}$$

$$\begin{aligned} \text{(b)} \quad \Pr_{\lambda, X}(\lambda, k) &= \frac{f(\lambda, k)}{\Pr(X=k)} \\ &= \frac{\Pr(k|\lambda) f(\lambda)}{\Pr(X=k)} \end{aligned}$$

$$\begin{aligned} \therefore f_{\lambda|X}(\lambda|k) &= \frac{P_{X|\lambda}(k|\lambda) P(\lambda)}{P(k)} \\ &= \frac{e^{-\lambda} \lambda^k \cdot e^{-\theta \lambda}}{\left(\frac{\theta}{1+\theta}\right) \left(\frac{1}{1+\theta}\right)^k} \end{aligned}$$

$$\therefore f_{\lambda|X}(\lambda|k) = \frac{(1+\theta)^{k+1} \lambda^k e^{-(1+\theta)\lambda}}{k!}$$

$\therefore f_{\lambda|X}(\lambda|k) = \frac{(1+\theta)^{k+1} \lambda^k e^{-(1+\theta)\lambda}}{\Gamma(k+1)}$   
 which is the gamma prob. density  $f_{\lambda}$  with parameters  $(k+1, 1+\theta)$

Remember  
 $X \sim \text{gamma}(n, \lambda)$   
 $\Rightarrow f(x) = \frac{\lambda^n}{\Gamma(n)} x^{n-1} e^{-\lambda x}$   
 $x \geq 0$   
 where  $\Gamma(n) = (n-1)!$   
 $n = 1, 2, \dots$

Note that  
 $x \rightarrow \lambda$   
 $\lambda \rightarrow 1+\theta$   
 $n \rightarrow k+1$  — For gamma pdf

Pb 2.4.7 p.71

Suppose that  $X$  and  $Y$  are independent Random Variables, each having the same exponential distribution with parameter  $\alpha$ . What is the conditional probability density function for  $X$ , given that  $Z = X + Y = z$ ?

Ans:  $f_{X|Z}(x|z) = \frac{f_{X,Y}(x,y)}{f(z)} \quad (1)$

$$\begin{aligned} \therefore f_{X,Y}(x,y) &= f_X(x) f_Y(y) \\ &= \alpha^2 e^{-\alpha(x+y)} \end{aligned}$$

$\therefore X \sim \text{exp}(\alpha), Y \sim \text{exp}(\alpha)$   
 and  $Z = X + Y$   
 $\therefore Z \sim \text{gamma}(2, \alpha)$

$\therefore f_{X,Y}(x,y) = \alpha^2 e^{-\alpha z}$   
 which equals the density  $f_{X,Z}(x,z)$  for  $0 \leq x \leq z$

$$\therefore f(z) = \frac{\alpha^2}{\Gamma(2)} z e^{-\alpha z}$$

$$\therefore f_{X,Z}(x,z) = \alpha^2 e^{-\alpha z} \quad (3)$$

$$\therefore f(z) = \alpha^2 z e^{-\alpha z} \quad (2)$$

Substitute (2) and (3) in (1), we get  
 $f_{X|Z}(x|z) = \frac{\alpha^2 e^{-\alpha z}}{\alpha^2 z e^{-\alpha z}} = \frac{1}{z}, 0 \leq x \leq z$

$\therefore X$  and  $Y$  are independent r.v.s  
 $\therefore f_{X,Y}(x,y) = f_X(x) f_Y(y)$

$\therefore X|Z$  is uniformly distributed over  $[0, z]$ .  $\#$