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# Lecture 10 Random Sums

p. 59 Textbook

Theorem: the moments of a random sum

$$E(\delta_k) = \mu, \quad \text{Var}(\delta_k) = \sigma^2$$

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

$\mu = \frac{\sigma^2}{\tau^2}$   
 $\sigma^2 = \tau^2 \mu$

Then  $E(X) = \mu\nu$

$X = \delta_1 + \delta_2 + \dots + \delta_k + \dots + \delta_N$

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

proof The random sum is

$$X = \begin{cases} 0 & , N=0 \\ \delta_1 + \delta_2 + \dots + \delta_N & , N>0 \end{cases}$$

i) To prove that  $E(X) = \mu\nu$

$$E(X) = \sum_{n=0}^{\infty} E[X | N=n] P_N(n)$$

$E[g(X)] = \sum_{n=0}^{\infty} E[g(X) | N=n] P_N(n)$

$\therefore E(X) = \sum_{n=1}^{\infty} E[\delta_1 + \delta_2 + \dots + \delta_N | N=n] P_N(n)$   
Defn of X

$$\Rightarrow E(X) = \sum_{n=1}^{\infty} E[\xi_1 + \xi_2 + \dots + \xi_n] P_N(n)$$

where  $N$  is independent of  $\xi_1, \xi_2, \dots$

$$\therefore E(\xi_k) = \mu, k=1, 2, \dots, n$$

Remember  
For Independent  
 $A, B$   
 $Pr(A|B) = Pr(A)$   
 $\Rightarrow E(A|B) = E(A)$

$$\therefore E(X) = \sum_{n=1}^{\infty} n \mu P_N(n)$$

$$= \mu \sum_{n=1}^{\infty} n P_N(n)$$

$$\therefore E(X) = \mu E(N) = \mu \sigma$$

ii) To prove that  $Var(X) = \sigma^2 + \mu^2 \sigma^2$

$$Var(X) = E[(X - \mu \sigma)^2]$$

$$= E[X - N\mu + N\mu - \sigma\mu]^2$$

$$\therefore Var(X) = E[(X - N\mu)^2] + E[\mu^2(N - \sigma)^2] + 2E[\mu(X - N\mu)(N - \sigma)]$$

$$\begin{aligned} E[(X - N\mu)^2] &= \sum_{n=0}^{\infty} E[(X - N\mu)^2 | N=n] P_N(n) \\ &= \sum_{n=1}^{\infty} E[(\xi_1 + \xi_2 + \dots + \xi_n - n\mu)^2 | N=n] P_N(n) \\ &= \sum_{n=1}^{\infty} E(\xi_1 + \xi_2 + \dots + \xi_n - n\mu)^2 \cdot P_N(n) \end{aligned}$$

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$$\therefore \text{Var}(\sum_k) = \sigma^2, k=1, 2, \dots, n$$

$$\text{Var}(\sum_k) = E(\sum_k - \mu)^2 = \sigma^2$$

$$\therefore E[(X - N\mu)^2] = \sum_{n=1}^{\infty} n \sigma^2 P_N(n)$$

$$= \sigma^2 \sum_{n=1}^{\infty} n P_N(n)$$

$$\therefore E[(X - N\mu)^2] = \boxed{\sigma^2} \text{ where } \sum_{n=1}^{\infty} n P_N(n) = \sigma$$

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$$\begin{aligned} \text{So } E[\mu^2(N - \sigma)^2] &= \mu^2 E[(N - \sigma)^2] \\ &= \mu^2 \text{Var}(N) = \mu^2 \sigma^2 \end{aligned}$$

$$\begin{aligned} \text{Also } E[\mu(X - N\mu)(N - \sigma)] &= \mu \sum_{n=1}^{\infty} E[(X - n\mu)(n - \sigma) | N=n] P_N(n) \\ &= \mu \sum_{n=1}^{\infty} (n - \sigma) E[(X - n\mu) | N=n] P_N(n) \end{aligned}$$

$$= 0$$

where  $E[(X - n\mu) | N=n] = E(X - n\mu)$   
independent

$$\begin{aligned} &= E(\sum_1 + \sum_2 + \dots + \sum_n - n\mu) \\ &= n\mu - n\mu = 0 \end{aligned}$$

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Substitute (2), (3) and (4) in (1), we get

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

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Pb 2.3.1 Text book

$N \sim \text{poisson}(\lambda)$ ,  $\sum_k \sim \text{Bernoulli}(p)$   
 $\Rightarrow E(N) = \mu = \lambda$   
 $\text{Var}(N) = \tau^2 = \lambda$   
 $\Rightarrow E(\sum_k) = \mu = p$   
 $\text{Var}(\sum_k) = \sigma^2 = p(1-p)$

For  $Z = \sum_1 + \sum_2 + \dots + \sum_N$ ,  $N > 0$   
 $\boxed{E(Z) = \mu \sigma = \lambda p}$

$$\boxed{\text{Var}(Z) = \sigma \sigma^2 + \mu^2 \tau^2}$$
$$= \lambda p(1-p) + p^2 \lambda$$

$$\therefore \text{Var}(Z) = \lambda p - \lambda p^2 + p^2 \lambda$$

$$\therefore \boxed{E(Z) = \lambda p}, \quad \boxed{\text{Var}(Z) = \lambda p}$$

$\therefore Z \sim \text{poisson}(\lambda p)$

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