

M339J: April 7<sup>th</sup>, 2021.

Practice 2: Problem 2.11.

$$X | \Lambda \sim U(0, \Lambda)$$

$$\Lambda \sim U(0, c)$$

$$c = 10$$

$$\mathbb{P}[X \leq 3] = ?$$

$$\mathbb{P}[X \leq 3] = F_X(3)$$

For a mixing dist'n:

$$F_X(x) = \int_0^{10} F_{X|\Lambda}(x|\lambda) f_{\Lambda}(\lambda) d\lambda$$

$= \frac{1}{10}$

$$\begin{cases} 0 & \text{for } x \leq 0 \\ \frac{x}{\lambda} & \text{for } 0 < x < \lambda \\ 1 & \text{for } x \geq \lambda \end{cases}$$

$$F_X(3) = \int_0^{10} F_{X|\Lambda}(3|\lambda) f_{\Lambda}(\lambda) d\lambda$$

$$= \int_0^3 1 \cdot f_{\Lambda}(\lambda) d\lambda + \int_3^{10} \left(\frac{3}{\lambda}\right) \cdot f_{\Lambda}(\lambda) d\lambda$$

$\leftarrow \lambda < 3 \Rightarrow F_{X|\Lambda}(3|\lambda) = 1$

$$= \int_0^3 \frac{1}{10} d\lambda + \int_3^{10} \frac{3}{\lambda} \cdot \frac{1}{10} d\lambda$$

$$= \frac{3}{10} + \frac{3}{10} \int_3^{10} \frac{1}{\lambda} d\lambda = \frac{3}{10} + \frac{3}{10} \ln\left(\frac{10}{3}\right) = \dots$$

## Binomial Distribution.

Consider  $m$  independent, identically dist'd risks w/ probability  $q$  of making a claim

Formally, for  $j=1, 2, \dots, m$ , we set

$$I_j = \begin{cases} 1 & \text{if risk } j \text{ makes a claim} \\ 0 & \text{if not} \end{cases}$$

Then, for every  $j=1..m$ ,  $I_j \sim \text{Bernoulli}(q)$

and  $\{I_j, j=1..m\}$  are independent.

$N$ ... the number of claims made

$$\Rightarrow N = I_1 + I_2 + \dots + I_m = \sum_{j=1}^m I_j$$

Q: What is the pmf of  $N$ ?

Note: There are many ways to answer this question; I chose this approach to illustrate a particular method.

→: We want to get and then use the pgf of  $N$ !

Start w/  $\{I_j, j=1..m\}$  which are independent, identically dist'd.

$$P_N(z) = \prod_{j=1}^m P_{I_j}(z) = (P_{I_1}(z))^m$$

independence  $I_j, j=1..m$  are identically dist'd

For a single Bernoulli trial:

$$P_{I_1}(z) = \mathbb{E}[z^{I_1}] = p_{I_1}(0) z^0 + p_{I_1}(1) z^1 = (1-q) + q \cdot z$$

by the def'n of pgf

$\Rightarrow$

$$P_{I_1}(z) = 1 + q(z-1)$$

$\Rightarrow$

$$P_N(z) = (1 + q(z-1))^m$$

$$P_N(z) = ((1-q) + q \cdot z)^m$$

Use the binomial formula

$$P_N(k) = \binom{m}{k} (1-q)^{m-k} \cdot q^k, \quad k=0,1,\dots,m$$

Q:  $E[N] = ?$

→:  $E[N] = E[I_1 + I_2 + \dots + I_m] =$  ← linearity of  $E$

$$= E[I_1] + E[I_2] + \dots + E[I_m] = m \cdot q$$

Q:  $\text{Var}[N] = ?$

→:  $\text{Var}[N] = \text{Var}[I_1 + \dots + I_m] =$  ← independence

$$= \text{Var}[I_1] + \dots + \text{Var}[I_m] = m \cdot q(1-q)$$

w/  $\text{Var}[I_1] = E[I_1^2] - (E[I_1])^2 = q - q^2 = q(1-q)$

Note on the "counting" distributions:

	<u>mean</u>		<u>variance</u>
Poisson( $\lambda$ )	$\lambda$	=	$\lambda$
NegBinomial( $r, \beta$ )	$r\beta$	<	$r\beta(1+\beta)$
Binomial( $m, q$ )	$mq$	>	$mq(1-q)$

Example. Let  $N \sim \text{Poisson}(\lambda)$  be our frequency r.v.  
Every loss is from:

- Category 1 w/ probab.  $p_1$
- Category 2 w/ probab.  $p_2$  w/  $p_1 + p_2 = 1$

Let  $N_i$  be the number of losses from Category  $i$ ,  $i=1,2$ .  
From our "thinning" theorem, we know that

$$N_i \sim \text{Poisson}(\lambda_i = p_i \cdot \lambda) \quad i=1,2.$$

We also know that  $N_1$  and  $N_2$  are independent.

Q: Given that  $N=m$ , what is the probability that  $N_1=k$ , for  $k=0,1,\dots,m$ ?

→:

$$\mathbb{P}[N_1=k \mid N=m] = \left( \text{by the def'n of conditional probability} \right)$$

$$= \frac{\mathbb{P}[N_1=k, N=m]}{\mathbb{P}[N=m]} =$$

$$= \frac{\mathbb{P}[N_1=k, N_1+N_2=m]}{\mathbb{P}[N=m]}$$

$$= \frac{\mathbb{P}[N_1=k, N_2=m-k]}{\mathbb{P}[N=m]}$$

(We know that  $N_1$  &  $N_2$  are independent)

Please, complete the calculation and recognize the conditional dist'n of  $N_1$  given that  $N=m$ . 😊