



**Answer the following questions.**

**Q1: [6+3+1]**

An automobile insurance policy with no coverage modifications has the following possible losses, with probabilities in parentheses: 100(0.4), 500(0.2), 1000(0.2), 2500(0.1), and 10000(0.1). Determine the probability mass functions and expected values for each of the following.

- The excess loss and left censored and shifted variables, where the deductible is set at 750.
- The limited loss variable with a limit of 750.
- Show that the sum of the expected values of the limited loss and left censored and shifted random variables is equal to the expected value of the original random variable.

**Q2: [3+3]**

- Obtain the mgf and pgf for the Poisson distribution.
- Demonstrate that the Weibull distribution is a scale distribution.

**Q3: [3+3+3]**

Let  $X$  have cdf  $F_X(x) = 1 - \left(\frac{\theta}{x + \theta}\right)^\alpha$ . Determine the cdf of the inverse, transformed, and inverse transformed distributions. Clarify the names of distributions.

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## The Model Answer

**Q1: [6+3+1]**

(a)

$x_j$	100	500	1000	2500	10000
$p(x_j)$	0.4	0.2	0.2	0.1	0.1

$$\begin{aligned} S(d) &= S(750) \\ &= \Pr(X > 750) \\ &= 0.4 \end{aligned}$$

For the mean of excess loss variable

$x_j - d$	250	1750	9250
$\frac{p(x_j)}{S(d)}$	0.5	0.25	0.25

$$\begin{aligned} \therefore e_X(d) &= 250(0.5) + 1750(0.25) + 9250(0.25) \\ &= 2875 \end{aligned}$$

For the mean of left censored and shifted variable

$$E(Y^L) = E[(X - d)_+]$$

$x_j - d$	0	250	1750	9250
$p(x_j)$	0.6	0.2	0.1	0.1

$$\begin{aligned} \therefore E(Y^L) &= 0(0.6) + 250(0.2) + 1750(0.1) + 9250(0.1) \\ &= 1150 \end{aligned}$$

(b)

For the mean of limited loss variable

$x_j \wedge d$	100	500	750
$p(x_j)$	0.4	0.2	0.4

$$\begin{aligned} E(X \wedge d) &= 100(0.4) + 500(0.2) + 750(0.4) \\ &= 440 \end{aligned}$$

(c)

$$\begin{aligned} E(X) &= 100(0.4) + 500(0.2) + 1000(0.2) + 2500(0.1) + 10000(0.1) \\ &= 1590 \end{aligned}$$

Clearly,  $E(X) = E[(X - d)_+] + E(X \wedge d)$

**Q2: [3+3]**

(a)

The pgf is

$$\begin{aligned}
P_X(z) &= \sum_{x=0}^{\infty} z^x \frac{\lambda^x e^{-\lambda}}{x!} \\
&= e^{-\lambda} \sum_{x=0}^{\infty} \frac{(z\lambda)^x}{x!} \\
&= e^{-\lambda} e^{z\lambda}
\end{aligned}$$

$$\therefore P_X(z) = e^{\lambda(z-1)}$$

The mgf is

$$M_X(z) = P_X(e^z)$$

$$\therefore M_X(z) = \exp[\lambda(e^z - 1)]$$

(b)

For the Weibull distribution,  $X \sim \text{Weibull}(\theta, \tau)$

and the distribution function is  $F_X(x) = 1 - e^{-(x/\theta)^\tau}$

let  $Y = cX$ ,  $c > 0$ , then

$$\begin{aligned}
F_Y(y) &= pr(Y \leq y) \\
&= pr(X \leq \frac{y}{c})
\end{aligned}$$

$$\therefore F_Y(y) = 1 - e^{-(y/c\theta)^\tau}$$

which is a Weibull distribution with parameters  $\tau$  and  $c\theta$

$\therefore \theta$  is a scale parameter.

$\therefore$  The Weibull distribution is a scale distribution

**Q3: [3+3+3]**

For Pareto  $(\alpha, \theta)$  distribution,  $F_X(x) = 1 - \left(\frac{\theta}{x + \theta}\right)^\alpha$ .

So, we could obtain the following:

(1) The inverse distribution has cdf

$$\begin{aligned}
F_Y(y) &= 1 - F_X(y^{-1}), \quad \tau = -1 \\
&= \left(\frac{\theta}{y^{-1} + \theta}\right)^\alpha
\end{aligned}$$

$$\therefore F_Y(y) = \left(\frac{y}{y + \theta^{-1}}\right)^\alpha$$

which is the inverse Pareto  $(\alpha, \theta^{-1})$  distribution.

(2) The transformed distribution has cdf

$$F_Y(y) = F_X(y^\tau), \quad \tau > 0$$

$$= 1 - \left(\frac{\theta}{y^\tau + \theta}\right)^\alpha$$

$$\therefore F_Y(y) = 1 - \left( \frac{1}{1 + (y / \theta^{1/\tau})^\tau} \right)^\alpha$$

which is the Burr  $(\alpha, \theta^{1/\tau}, \tau)$  distribution.

(3) The inverse transformed distribution has cdf

$$F_Y(y) = 1 - F_X(y^{-\tau}) \quad \text{Theorem for negative } \tau$$

$$= 1 - \left[ 1 - \left( \frac{\theta}{\theta + y^{-\tau}} \right)^\alpha \right]$$

$$= \left( \frac{\theta}{\theta + y^{-\tau}} \right)^\alpha$$

$$= \left( \frac{y^\tau}{y^\tau + \theta^{-1}} \right)^\alpha$$

$$= \left( \frac{y^\tau}{y^\tau + (\theta^{-1/\tau})^\tau} \right)^\alpha$$

$$\therefore F_Y(y) = \left( \frac{(y / \theta^{-1/\tau})^\tau}{1 + (y / \theta^{-1/\tau})^\tau} \right)^\alpha$$

which is the inverse Burr  $(\alpha, \theta^{-1/\tau}, \tau)$  distribution.

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