

Stat510(Section001): Homework #3

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Problem 1

we know that the mean of e^{-x} is 1, the variance of it is also 1. Hence, we get the skewness:

$$E\left(\frac{(x-1)^3}{\sigma^3}\right) = \int_0^\infty (x-1)^3 e^{-x} dx = \int_0^\infty (x^3 - 3x^2 + 3x - 1)e^{-x}$$

We have:

$$\begin{aligned} \int_0^\infty x e^{-x} dx &= (-x e^{-x} + e^{-x}) \Big|_{x=0}^\infty = 1 \\ \int_0^\infty x^2 e^{-x} dx &= (-x^2 e^{-x}) \Big|_{x=0}^\infty + 2 \int_0^\infty x e^{-x} dx = 2 \\ \int_0^\infty x^3 e^{-x} dx &= (-x^3 e^{-x}) \Big|_{x=0}^\infty + 3 \int_0^\infty x^2 e^{-x} dx = 6 \end{aligned}$$

Hence:

$$E\left(\frac{(x-1)^3}{\sigma^3}\right) = 6 - 6 + 3 - 1 = 2$$

The skewness of e^{-x} is 2.

Problem 2

We can know that:

$$Y = \tan(x)d$$

Hence:

$$P(Y \leq y) = P(\tan(x)d \leq y) = P(X \leq \arctan(\frac{y}{d})) = \frac{\arctan(y/d)}{\pi/2} = \frac{2\arctan(y/d)}{\pi}$$

Hence:

$$f_Y(y) = F'_Y(y) = \frac{2}{\pi} \frac{d^2}{d^2 + y^2}$$

is the pdf of Y.

$$E(Y) = \frac{2}{\pi} \int_{y=0}^\infty \frac{d^2 y}{d^2 + y^2} dy = \frac{2}{\pi} \frac{d^2}{2} \log(d^2 + y^2) \Big|_{y=0}^\infty$$

We can see that $E(Y)$ does not exist.

Problem 3

Part a

Proof. We get:

$$E(|x-a|) = \int_{x=-\infty}^\infty |x-a| f_X(x) dx = \int_{x=-\infty}^a (a-x) f_X(x) dx + \int_{x=a}^\infty (x-a) f_X(x) dx$$

To get the min of this function. We calculate $\frac{d(E(|x-a|))}{d(a)} = 0$:

$$\frac{d(E(|x-a|))}{d(a)} = (a-x) f_X(a) \Big|_{x=a} + \int_{x=-\infty}^a f_X(x) dx + (x-a) f_X(a) \Big|_{x=a} - \int_{x=a}^\infty f_X(x) dx = 0$$

Thus,

$$\int_{x=-\infty}^a f_X(x) dx = \int_{x=a}^\infty f_X(x) dx \rightarrow F_X(a) = 1 - F_X(a)$$

We have:

$$F_X(a) = \frac{1}{2} \rightarrow P(X \leq a) = \frac{1}{2}$$

Thus, we prove that $E(|x - a|)$ is minimized at median. \square

Part b

Proof. We have:

$$\begin{aligned} &= \int_{x=-\infty}^{\infty} (x - m)^2 f_X(x) dx - \int_{x=-\infty}^{\infty} (x - E(x))^2 f_X(x) dx \\ &= -2m \int_{x=-\infty}^{\infty} x f_X(x) dx + m^2 \int_{x=-\infty}^{\infty} f_X(x) dx + 2E(x) \int_{x=-\infty}^{\infty} x f_X(x) dx - E(x)^2 \int_{x=-\infty}^{\infty} f_X(x) dx \\ &= -2mE(x) + m^2 + 2E(x)^2 - E(x)^2 \\ &= 2E(x)(E(x) - m) + m^2 - E(x)^2 \end{aligned}$$

Now we need to prove this formula is greater than 0. We have:

$$\begin{aligned} 2E(x)(E(x) - m) + m^2 - E(x)^2 &= 2E(x)(E(x) - m) - (E(x)^2 - m^2) \\ &= 2E(x)(E(x) - m) - (E(x) - m)(E(x) + m) \\ &= (E(x) - m)(E(x) - m) \\ &= (E(x) - m)^2 \geq 0 \end{aligned}$$

Thus we prove the conclusion. \square

Problem 4

we have:

$$E(Y^n) = \int_{y=x}^{\infty} \frac{\alpha x^\alpha}{y^{\alpha+1}} y^n dy = \int_{y=x}^{\infty} \alpha x^\alpha y^{n-\alpha-1} dy$$

Now we have three situaion:

1. $n > \alpha$:

$$\begin{aligned} E(Y^n) &= \int_{y=x}^{\infty} \alpha x^\alpha y^{n-\alpha-1} dy \\ &= \frac{1}{n - \alpha} \alpha x^\alpha y^{n-\alpha} \Big|_{y=x}^{\infty} = \infty \end{aligned}$$

Hence, this situaion, the expection does not exists.

2. $n = \alpha$:

$$\begin{aligned} E(Y^n) &= \int_{y=x}^{\infty} \frac{\alpha x^\alpha}{y} dy \\ &= \alpha x^\alpha \log(y) \Big|_{y=x}^{\infty} = \infty \end{aligned}$$

This situaion, the expection does not exists.

3. $n < \alpha$:

$$\begin{aligned} E(Y^n) &= \int_{y=x}^{\infty} \alpha x^\alpha y^{n-\alpha-1} dy \\ &= \frac{1}{n - \alpha} \alpha x^\alpha y^{n-\alpha} \Big|_{y=x}^{\infty} = \frac{\alpha x^n}{\alpha - n} \end{aligned}$$

Hence, when $n < \alpha$, we have $E(Y^n)$ exists. Then we consider the skewness. If the skewness exists, then $E(Y^2)$ must exist. Hence $\alpha > 2$. And also $E(Y - \frac{\alpha x}{\alpha-1})^3$ need to exist. We have:

$$\begin{aligned} E(Y - \frac{\alpha x}{\alpha-1})^3 &= \int_{y=x}^{\infty} \frac{\alpha x^\alpha}{y^{\alpha+1}} (Y - E(y))^3 dy \\ &= E(Y^3) - 3E(y)E(Y^2) + 2(E(y))^3 \\ &= \frac{\alpha}{\alpha-3}x^3 - \frac{3\alpha^2}{(\alpha-1)(\alpha-2)}x^3 + \frac{2\alpha^3}{(\alpha-1)^3}x^3 \end{aligned}$$

Thus, $\alpha > 3$, skewness exists, We also have:

$$Var(Y) = E(Y^2) - E(Y)^2 = \frac{\alpha}{\alpha-2}x^2 - \frac{\alpha^2}{(\alpha-1)^2}x^2$$

Hence:

$$skewness = \frac{E(Y - \frac{\alpha x}{\alpha-1})^3}{Var(Y)^{3/2}}$$

And, we have mgf:

$$E[e^{ty}] = \int_{y=x}^{\infty} \alpha x^\alpha y^{-\alpha-1} e^{ty} dy$$

we can know, for any n:

$$\lim_{y \rightarrow \infty} \frac{e^{ty}}{y^n} = \infty$$

Which shows that $y^{-\alpha-1}e^{ty}$ does not converge. Hence $E[e^{ty}] = \infty$. mgf does not exist.

Problem 5

Part i

We have:

$$\begin{aligned} E(\frac{1}{1+x}) &= \sum_{x=0}^n \frac{1}{1+x} \binom{n}{x} p^x (1-p)^{n-x} \\ &= \sum_{x=0}^n \frac{1}{1+x} \frac{n!}{(n-x)!x!} p^x (1-p)^{n-x} \\ &= \frac{1}{n+1} \sum_{x=0}^n \frac{(n+1)!}{(n-x)!(x+1)!} p^x (1-p)^{n-x} \\ &= \frac{1}{(n+1)p} \sum_{k=1}^n \frac{(n+1)!}{(n-x)!k!} p^k (1-p)^{n+1-k} \\ &= \frac{1}{(n+1)p} (1 - (1-p)^{n+1}) \end{aligned}$$

Part ii

$$\begin{aligned}
E\left(\frac{1}{1+x}\right) &= \sum_{x=0}^{\infty} \frac{1}{1+x} \frac{\lambda^x}{x!} e^{-\lambda} \\
&= \frac{e^{-\lambda}}{\lambda} \sum_{x=0}^{\infty} \frac{\lambda^{x+1}}{(x+1)!} \\
&= \frac{e^{-\lambda}}{\lambda} \sum_{k=1}^{\infty} \frac{\lambda^k}{k!} \\
&= \frac{e^{-\lambda}}{\lambda} (e^{\lambda} - 1) \\
&= \frac{1 - e^{-\lambda}}{\lambda}
\end{aligned}$$

Now let us consider the relationship between them. We know that, when n is large enough, Poisson can be viewed as a binomial distribution with parameter np . Now, let us consider:

$$\lim_{n \rightarrow \infty} \frac{1 - (1-p)^{n+1}}{(n+1)p} = \lim_{n \rightarrow \infty} \frac{1 - (1 - \frac{\lambda}{n})^{n+1}}{(n+1)p}$$

we have:

$$\begin{aligned}
\lim_{n \rightarrow \infty} (n+1)p &= \lambda \\
\lim_{n \rightarrow \infty} (1 - \frac{\lambda}{n})^{n+1} &= e^{-\lambda}
\end{aligned}$$

Hence:

$$\lim_{n \rightarrow \infty} \frac{1 - (1 - \frac{\lambda}{n})^{n+1}}{(n+1)p} = \frac{1 - e^{-\lambda}}{\lambda}$$

Which is to say, the relationship between Poisson and Binomial distribution is the same relationship between their $\frac{1}{1+x}$.

Problem 6

With l'Hopital rule, we can know:

$$\lim_{x \rightarrow \infty} e^{tx - \log(x)^2} \rightarrow \infty$$

That is to say, for any c , we have a k , for every $x > k$, we have:

$$e^{tx - \log(x)^2} > c$$

Hence:

$$\frac{1}{\sqrt{2\pi}} \int_{x=k}^{\infty} \frac{e^{tx - \log(x)^2}}{x} \geq \frac{1}{\sqrt{2\pi}} \int_{x=k}^{\infty} \frac{c}{x} = \infty$$

Hence:

$$\frac{1}{\sqrt{2\pi}} \left(\int_{x=0}^k \frac{e^{tx - \log(x)^2}}{x} + \int_{x=k}^{\infty} \frac{e^{tx - \log(x)^2}}{x} \right) = \infty$$

Thus, the mgf does not exist.

Problem 7

Part a

We let X be always earn 1 dollars, and Y be the gamble. We have:

$$E(U(X)) = 1, E(U(Y)) = 0.001 \times 500^c$$

if c is not even, we have:

$$0.001 \times 500^c > 1 \rightarrow c > \log_{500}(1000)$$

Hence:

$$c > \log_{500}(1000)$$

Part b

We let X be the earn Z dollars and Y be take the gamble. Then we have:

$$E(U(X)) = Z^c, E(U(Y)) = \sum_{T=1}^{\infty} (2^{T-1})^c \frac{1}{2}^T$$

We have:

$$E(U(Y)) = \sum_{T=1}^{\infty} 2^{(c-1)T-c}$$

if $c \geq 1$, we have $E(U(Y)) = \infty$, Hence we will always be prefer to take the gamble. Now we consider when $c < 1$ we have:

$$E(U(Y)) = \sum_{T=1}^{\infty} 2^{(c-1)T-c} = \frac{\frac{1}{2}}{1 - 2^{c-1}} = \frac{1}{2 - 2^c}$$

$$\frac{1}{2 - 2^c} \geq Z^c \rightarrow 2Z^c - (2Z)^c \leq 1$$

Hence if $c > 1$ or $2Z^c - (2Z)^c \leq 1$, we prefer to take the gamble.

Problem 8

Part a

We assume the X is an exponential distribution with the following pdf:

$$f_X(x) = \frac{1}{\beta} e^{-\frac{x}{\beta}} \quad x > 0$$

We let $\lambda = \frac{1}{\beta}$, we can get a new form of pdf:

$$f_X(x) = \lambda e^{-\lambda x} \quad x > 0$$

We will use this form to calculate. Let us consider the M(c) First:

$$M(c) = \lambda \int_0^{\infty} e^{(c-\lambda)x} dx = \frac{\lambda}{c-\lambda} e^{(c-\lambda)x} \Big|_0^{\infty}$$

if $c \geq \lambda$ (in other word), M(c) does not exist. and if $c < \lambda$ (in other word), we have:

$$M(c) = \frac{\lambda}{\lambda - c}$$

Thus, c must less than λ , then We have:

$$F_Y(y) = \frac{1}{M(c)} \left(\frac{\lambda}{\lambda - c} - \frac{\lambda}{\lambda - c} e^{-(\lambda - c)y} \right), y > 0$$

Hence:

$$f_Y(y) = F'_Y(y) = (\lambda - c)e^{-(\lambda - c)y}, y > 0$$

Hence Y is also an exponential distribution with parameter ' λ ' (in a form of $f_Y(y) = \lambda e^{-\lambda y}$) as $\lambda - c$. If we write it in a form of ' β ' way, we have:

$$f_Y(y) = \left(\frac{1}{\beta} - c\right)e^{-(\frac{1}{\beta} - c)y}, y > 0$$

Therefore, in a ' β ' way, Y is an exponential distribution with parameter ' β ' (in a form of $f_Y(y) = \frac{1}{\beta}e^{-\frac{y}{\beta}}$) as $\frac{\beta}{1 - \beta c}$.

Part b

We have:

$$f_Y(y) = F'_Y(y) = \frac{1}{M(c)}e^{cy}f_x(y)$$

And the mgf of Y is:

$$M_Y(a) = \frac{1}{M(c)} \int_{-\infty}^{\infty} e^{ay} e^{cy} f_x(y) dy = \frac{1}{M(c)} \int_{-\infty}^{\infty} e^{(a+c)y} f_x(y) dy = \frac{M(a+c)}{M(c)}$$

Hence, the mgf of Y has definition in a neighborhood of 0 if and only if the mgf of X is defined in a neighborhood of c.