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ASSIGNMENT 4  
MATH 356

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## QUESTION 1

**Part (a):** Let  $f_X = \frac{1}{2}e^{-|x|}$ . Then we have

$$\begin{aligned} M(t) &= \int_{\mathbb{R}} e^{tx} \left( \frac{1}{2} e^{-|x|} \right) dx \\ &= \frac{1}{2} \left( \int_{-\infty}^0 e^{tx+x} + \int_0^{\infty} e^{tx-x} \right) = \frac{1}{2} \left( \int_{-\infty}^0 e^{x(t+1)} + \int_0^{\infty} e^{x(t-1)} \right) \\ &= \frac{1}{2} \left( \frac{1}{t+1} e^{x(t+1)} \Big|_{-\infty}^0 + \frac{1}{t-1} e^{x(t-1)} \Big|_0^{\infty} \right) \end{aligned}$$

The first term is finite exactly when  $t > -1$ , and the second term is finite when  $t < 1$ . Thus, we require that  $|t| < 1$ . When this is true, we have:

$$\frac{1}{2} \left( \frac{1}{t+1} e^{x(t+1)} \Big|_{-\infty}^0 + \frac{1}{t-1} e^{x(t-1)} \Big|_0^{\infty} \right) = \frac{1}{2} \left( \frac{1}{t+1} - \frac{1}{t-1} \right) = \frac{1}{2} \left( \frac{-2}{t^2-1} \right) = \frac{1}{1-t^2}$$

Thus,  $M_X(t) = \frac{1}{1-t^2}$  for  $|t| < 1$ , and  $\infty$  otherwise.

**Part (b):** Suppose that  $|t| < 1$ . The derivatives of  $\frac{1}{1-t^2}$  will get hairy quick, so consider its Taylor expansion, or, rather that of  $\frac{1}{1-t}$ .

Since the derivatives of  $\frac{1}{1-t}$  are:

$$f' = \frac{1}{(1-t)^2} \quad f'' = \frac{2!}{(1-t)^3} \quad f''' = \frac{3!}{(1-t)^4} \quad \dots \quad f^{(n)} = \frac{n!}{(1-t)^{n+1}}$$

we have that  $f^{(n)}(0) = n!$ . Thus,  $\frac{1}{1-t}$  expanded about 0 is the series

$$1 + t + t^2 + t^3 + \dots = \sum_{n=0}^{\infty} t^n$$

and with  $t \rightarrow t^2$  we have that  $\frac{1}{1-t^2} = \sum_{n=0}^{\infty} t^{2n}$ . Evaluating at  $t = 0$ , notice that, when the exponential of  $t$  is 0, we yield the one and only non-zero term in the series. The  $k$ -th derivative of this final Taylor series is as follows:

$$f^{(k)} = \sum_{n=0}^{\infty} 2n(2n-1)(2n-2)\dots(2n-k+1)t^{2n-k}$$

For the first term: if  $t = -1$ ,  $\frac{1}{t+1}$  diverges, and if  $t < -1$ , then  $(t+1)$  is negative, and thus  $e^{x(t+1)}$  will evaluate to infinity at the lower bound.

For the second term: let  $t = 1$ . Then  $\frac{1}{t-1}$  is divergent. And if  $t > 1$ , we have that  $(t-1)$  is positive, and thus  $e^{x(t-1)}$  will evaluate to infinity at our upper bound.

...as we “pull down” the power of each subsequent derivative. For the  $t$  term to be non-zero when evaluated at 0, we require that its power be 0, or  $n = \frac{k}{2}$ , which only occurs at even derivatives. In fact, if  $k$  is odd, then  $t^{2n-k}$  will be odd, and thus evaluate to 0 for all powers. Alternatively, the  $\frac{k}{2}$ -th term will simply be  $k(k-1)(k-2)\dots 1 = k!$

In short, we have that  $M^{(k)}(0)$  is non-zero for *even*  $k$ , where  $M^{(k)}(0) = k!$ . Otherwise  $M^{(k)}(0) = 0$ .

We can finally express the  $n$ -th moment of  $X$  as

$$\mathbb{E}[X^n] = M^{(n)}(0) = \begin{cases} n! & \text{when } n \text{ is even} \\ 0 & \text{when } n \text{ is odd} \end{cases}$$

## QUESTION 2

**Part (a):** Let  $X \sim \mathcal{N}(0, 1)$ , and let  $Y := e^X$ . When  $y \leq 0$ ,  $\mathbb{P}(Y \leq y) = \mathbb{P}(e^X \leq y) = 0$ , since  $e^X$  does not span  $\mathbb{R}^-$ . For a given  $y > 0$ , the set of all  $\{x : e^x = y\}$  is just  $\ln(y)$ , as  $e^x$  is a one-to-one mapping. Further, note that  $g'(x) = e^{\ln(y)} = y$  is zero nowhere when  $y > 0$ , so we can apply the following formula:

$$f_Y(y) = \sum_{x: e^x=y} \frac{f_X}{|g'(x)|} = \frac{f_X}{|g'(x)|} \Big|_{x=\ln(y)} = \frac{e^{-\frac{\ln^2(y)}{2}}}{\sqrt{2\pi}y}$$

$$\text{since } |g'| = |y| = y \text{ and } f_X = \varphi(x) = \frac{e^{-\frac{x^2}{2}}}{\sqrt{2\pi}} \Big|_{x=\ln(y)} = \frac{e^{-\frac{\ln^2(y)}{2}}}{\sqrt{2\pi}}.$$

$$\text{In total, we have } f_Y(y) = \begin{cases} 0 & y \leq 0 \\ \frac{e^{-\frac{\ln^2(y)}{2}}}{\sqrt{2\pi}y} & y > 0 \end{cases}$$

**Part (b):** We've seen from lecture that  $M_X(t) = e^{\frac{t^2}{2}}$  for  $X \sim \mathcal{N}(0, 1)$ . In other words,  $\mathbb{E}[e^{tX}] = e^{\frac{t^2}{2}}$ .

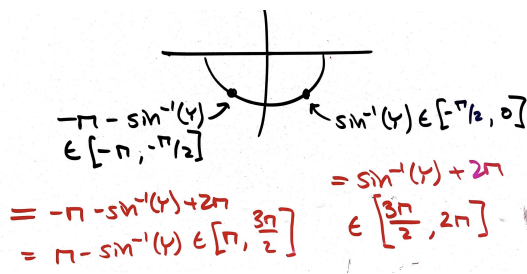
We have that  $\mathbb{E}[Y^n]$  is the  $n$ -th moment of  $Y$ , and this is  $\mathbb{E}[(e^X)^n] = \mathbb{E}[e^{nX}] = e^{\frac{n^2}{2}}$  from above.

$$\implies \mathbb{E}[Y^n] = e^{\frac{n^2}{2}}$$

### QUESTION 3

Let  $X \sim \text{Unif}[-\pi, 2\pi]$  and  $Y = \sin(X)$ . Define the function  $\sin^{-1}(y)$  on the interval  $[-\frac{\pi}{2}, \frac{\pi}{2}]$ , as usual. We have that  $f_X = \frac{1}{2\pi - (-\pi)} = \frac{1}{3\pi}$ . Recall:

$$\star f_Y = \sum \frac{f_X(x)}{|g'(x)|} \text{ over all } x \text{ such that } g(x) = y \text{ and } g'(x) \neq 0$$



When  $|y| > 1$ , we have immediately that  $\sin(x) \neq y$  for any  $x$ , and thus  $f_Y = 0$  for these values. Suppose now that  $y \in [-1, 0]$ . On the interval  $[-\pi, 2\pi]$ , one sees that  $\sin^{-1}(y) = -\pi - \sin^{-1}(y) = \pi - \sin^{-1}(y) = \sin^{-1}(y) + 2\pi$ .

Thus, when  $y \in [-1, 0]$ , there exist 4 equivalent solutions to  $g(x) = y$  on the interval  $x \in [-\pi, 2\pi]$ . From  $\star$ , we have

$$f_Y = \frac{1}{3\pi} \left( \frac{1}{|\cos(\sin^{-1}(y))|} + \frac{1}{|\cos(-\pi - \sin^{-1}(y))|} + \frac{1}{|\cos(\pi - \sin^{-1}(y))|} + \frac{1}{|\cos(\sin^{-1}(y) + 2\pi)|} \right) = \frac{4}{3\pi\sqrt{1-y^2}}$$

The following calculations led to this answer: first, we know that all the phase shifts given for  $\sin^{-1}$  are equivalent, so the expression is really  $\frac{4}{3\pi} \left( \frac{1}{|\cos(\sin^{-1}(y))|} \right)$ . Then, for a unit triangle,  $\cos(\sin^{-1}(y))$  is  $\frac{a}{1}$ , where  $\sin^{-1}(y)$  is the angle  $\theta$  at which  $\sin(\theta) = \frac{y}{1}$ , i.e. the angle adjacent to  $a$ . Thus, since  $a^2 + y^2 = 1$ , we have  $a = \cos(\sin^{-1}(y)) = \sqrt{1-y^2}$ . We conclude that  $f_Y = \frac{4}{3\pi\sqrt{1-y^2}}$  for  $y \in [-1, 0]$ .

Now consider  $y \in (0, 1]$ . We have that, for  $\sin^{-1}(y) : \mathbb{R}^+ \rightarrow (0, \frac{\pi}{2}]$ ,  $\sin^{-1}(y) = \pi - \sin^{-1}(y)$ , where the latter term has range  $[\frac{\pi}{2}, \pi)$ . Notice, however, that either of these expressions  $\pm 2\pi$  exits our allowed domain of  $[-\pi, 2\pi]$ . Thus, we have the only two  $x : g(x) = y$ , and by  $\star$ :

$$f_Y = \frac{1}{3\pi} \left( \frac{1}{|\cos(\sin^{-1}(y))|} + \frac{1}{|\cos(\pi - \sin^{-1}(y))|} \right) = \frac{2}{3\pi\sqrt{1-y^2}} \text{ as above}$$

Thus, the complete PDF of  $Y$  is the following:

$$f_Y(y) = \begin{cases} \frac{4}{3\pi\sqrt{1-y^2}} & y \in [-1, 0] \\ \frac{2}{3\pi\sqrt{1-y^2}} & y \in (0, 1] \\ 0 & \text{otherwise} \end{cases}$$

The "picture" is sin reflected over the positive y-axis.

## QUESTION 4

Consider the standard deviation  $\sigma = \sqrt{\text{Var}(X)}$ . This is defined in  $\mathbb{R}$  for all random variables. Thus,  $\text{Var}(X) \geq 0$ , or  $\mathbb{E}[X^2] - (\mathbb{E}[X])^2 \geq 0 \implies \mathbb{E}[X^2] \geq (\mathbb{E}[X])^2$  for all variables  $X$ .

Let  $M_X(1) = 3$  and  $M_X(2) = 4$ . Then  $\mathbb{E}[e^X] = 3$  and  $\mathbb{E}[e^{2X}] = 4$ . Define  $Y := e^X$ . This is well defined for any variable  $X$ .

We then have that  $\mathbb{E}[Y] = 3 \implies (\mathbb{E}[Y])^2 = 9$  and  $\mathbb{E}[Y^2] = 4$ .

Thus,  $\mathbb{E}[Y^2] < (\mathbb{E}[Y])^2$ , and we are done.

QUESTION 5

$$\text{Let } \mathbb{P}(X \leq x) = \begin{cases} 0 & x < 0 \\ x & 0 \leq x < \frac{1}{2} \\ 1 & x \geq \frac{1}{2} \end{cases}$$

Considering, alone, the “continuous part” of  $F_X$ , we have  $f_X = \frac{d}{dx}x = 1$  for  $0 \leq x < \frac{1}{2}$ . For  $x < 0$ , it’s clear to see that  $\rho_X(x) = 0$ , the CMF of the “discrete part.”

$F_X$  implies that the sum of probabilities over  $x \in [0, \frac{1}{2}]$  is 1. The continuous part contributes  $\int_0^{1/2} dx = \frac{1}{2}$ , and thus  $\mathbb{P}(X = \frac{1}{2}) = 1 - \frac{1}{2} = \frac{1}{2}$ .

Consider  $\mathbb{E}[e^{tX}]$ . For  $x$  defined on  $[0, \frac{1}{2})$ , this is

$$\int_0^{1/2} e^{tx} dx = \left. \frac{1}{t} e^{tx} \right|_0^{1/2} = \frac{e^{t/2} - 1}{t}$$

When  $x = \frac{1}{2}$ , we have  $\mathbb{E}[e^{tX}] = \sum_{x=1/2} e^{tx} \mathbb{P}(X = x) = \frac{e^{t/2}}{2}$

Over the entire domain of  $X$ , then, we have

$$\mathbb{E}[e^{tX}] = \frac{e^{t/2} - 1}{t} + \frac{e^{t/2}}{2}$$



## QUESTION 6

Consider  $X \sim \text{Geom}(p)$  and  $Y \sim \text{Poi}(\lambda)$ . We have

$$\mathbb{P}(X = k) = (1 - p)^{k-1} p \quad \forall k \geq 1 \quad \text{and} \quad \mathbb{P}(Y = k) = \frac{e^{-\lambda} \lambda^k}{k!} \quad \forall k \geq 0$$

Now let  $Y \rightarrow Y + 1$  be an affine transformation.  $f_{Y+1} = (1)f_Y(k-1) = \frac{e^{-\lambda} \lambda^{k-1}}{(k-1)!} \quad \forall k \geq 1$ . Then we have that  $\mathbb{P}(Y + 1 = X)$  is the sum of probabilities that  $\mathbb{P}(X = k)$  and  $\mathbb{P}(Y + 1 = k)$  for all valid  $k$ , as  $X$  and  $Y$  are independent.

$$\begin{aligned} \mathbb{P}(Y + 1 = X) &= \sum_{k \geq 1} \frac{(1 - p)^{k-1} p e^{-\lambda} \lambda^{k-1}}{(k-1)!} \\ &= p e^{-\lambda} \sum_{k \geq 1} \frac{(\lambda - \lambda p)^{k-1}}{(k-1)!} = p e^{-\lambda} \sum_{k \geq 0} \frac{(\lambda - \lambda p)^k}{k!} \\ &= e^{\lambda - p\lambda} p e^{-\lambda} = e^{-p\lambda} p \end{aligned}$$

QUESTION 7

Suppose we sample with replacement from an urn of 4 green, 3 yellow, and 2 white balls. Let  $N := \{\# \text{ of draws until one sees a green or yellow}\}$ . Since we're sampling without replacement, the probability that we pick a green or yellow is constant  $\frac{7}{9}$ . Thus,  $N$  is a geometric r.v. with

$$\mathbb{P}(N = k_1) = \left(\frac{2}{9}\right)^{k_1-1} \left(\frac{7}{9}\right)$$

After one stops picking, there is  $\mathbb{P} = \frac{4}{7}$  that the ball is green and  $\mathbb{P} = \frac{3}{7}$  the ball is yellow. Thus,  $Y$  as defined in the question has probability

$$\mathbb{P}(Y = k_2) = \begin{cases} \frac{4}{7} & k_2 = 1 \\ \frac{3}{7} & k_2 = 2 \\ 0 & \text{otherwise} \end{cases}$$

Combining, the joint probability function is

$$\rho_{N,Y}(k_1, k_2) = \begin{cases} \left(\frac{2}{9}\right)^{k_1-1} \left(\frac{7}{9}\right) \left(\frac{4}{7}\right) & k_2 = 1 \\ \left(\frac{2}{9}\right)^{k_1-1} \left(\frac{7}{9}\right) \left(\frac{3}{7}\right) & k_2 = 2 \\ 0 & \text{otherwise} \end{cases}$$

For  $N$ , the marginal probability function is given by summing all possible values of  $Y$ , i.e.,  $Y = \{1, 2\}$ :

$$\rho_N(k_1) = \left(\frac{2}{9}\right)^{k_1-1} \left(\frac{7}{9}\right) \left(\frac{4}{7}\right) + \left(\frac{2}{9}\right)^{k_1-1} \left(\frac{7}{9}\right) \left(\frac{3}{7}\right) = \left(\frac{2}{9}\right)^{k_1-1} \left(\frac{7}{9}\right)$$

For  $Y$ , its MPF is given by summing over possible values of  $N$ , i.e.  $\mathbb{N}$ . Note that, as in its joint function, we need to separate this into cases for  $k_2$ :

$$\rho_Y(k_2) = \begin{cases} \frac{4}{7} \sum_{k_1 \geq 1} \left(\frac{2}{9}\right)^{k_1-1} \left(\frac{7}{9}\right) & k_2 = 1 \\ \frac{3}{7} \sum_{k_1 \geq 1} \left(\frac{2}{9}\right)^{k_1-1} \left(\frac{7}{9}\right) & k_2 = 2 \\ 0 & \text{otherwise} \end{cases} = \begin{cases} \frac{4}{7} & k_2 = 1 \\ \frac{3}{7} & k_2 = 2 \\ 0 & \text{otherwise} \end{cases}$$

Since  $\sum_{k_1 \geq 1} \left(\frac{2}{9}\right)^{k_1-1} \left(\frac{7}{9}\right) = \frac{7}{9} \left(\frac{1}{1-\frac{2}{9}}\right) = 1$  by geometric series.

Now consider the event  $\{N = k \text{ and } Y = k\}$ . From above, we can express  $\mathbb{P}(N = k, Y = k) = \rho_{N,Y}(k, k)$ .

When  $k \leq 0$  and  $k > 2$ , this probability is 0, since one cannot choose a ball before the game starts, and  $Y$  is not defined for  $k > 2$ . Thus,  $\rho_{N,Y}(k, k) = \rho_N(k)\rho_Y(k) = 0$  for these values.

When  $k = 1$ , we have  $\rho_{N,Y}(1, 1) = \left(\frac{2}{9}\right)^0 \left(\frac{7}{9}\right)\left(\frac{4}{7}\right) = \left(\frac{7}{9}\right)\left(\frac{4}{7}\right) = \rho_N(1)\rho_Y(1) = \left(\frac{7}{9}\right)\left(\frac{4}{7}\right)$  using the expressions defined above.

When  $k = 2$ , we have  $\rho_{N,Y}(2, 2) = \left(\frac{2}{9}\right)\left(\frac{7}{9}\right)\left(\frac{3}{7}\right) = \left(\frac{14}{81}\right)\left(\frac{3}{7}\right) = \rho_N(2)\rho_Y(2) = \left(\frac{14}{81}\right)\left(\frac{3}{7}\right)$  using the expressions defined above.

Thus, over all integers  $k$ ,  $\rho_{N,Y}(k, k) = \rho_N(k)\rho_Y(k)$ , and thus

$$\mathbb{P}(N = k, Y = k) = \mathbb{P}(N = k)\mathbb{P}(Y = k)$$

so  $N$  and  $Y$  are independent.

And yes, the PMF of  $Y$  makes sense, since we are essentially flipping a coin with bias  $p = \frac{3}{7}$ .