$$p_K(k) = \begin{cases} \frac{1}{2n+1} & \text{for } k \in \{-n, -n+1, \dots, n\} \\ 0 & \text{otherwise} \end{cases}$$

$$p_{|K|}(|k|) = \begin{cases} \frac{2}{2n+1} & \text{for } |k| \in \{1, \dots, n\} \\ \frac{1}{2n+1} & \text{for } |k| = 0 \\ 0 & \text{otherwise} \end{cases}$$

$$\begin{split} X &= a^{|K|} \quad a > 0 \\ Y &= \log(X) = |K| \log(a) \\ \text{PMF of Y:} \end{split}$$

$$p_Y(y) = \begin{cases} \frac{2}{2n+1} & \text{for } y \in \{\log(a), 2\log(a), \dots, n\log(a)\} \\ \frac{1}{2n+1} & \text{if } y = 0 \\ 0 & \text{otherwise} \end{cases}$$

Problem 2

(a)
$$\sum_{x=-3}^{3} = x^2/a = 1$$
$$\sum_{x=-3}^{3} = x^2 = a$$
$$a = 2(1^2 + 2^2 + 3^2) = 28$$
$$a = 28$$
$$E[X] = \sum_{x=-3}^{3} x^3/28$$
$$= (1/28)[(-1)^3 + (-2)^3 + (-3)^3 + 0 + 1^3 + 2^3 + 3^3] = 0$$
$$E[X] = 0$$

$$Z = (X - E[X])^{2}$$

$$= X^{2} - (2X)E[X] + (E[X])^{2}$$

$$= X^{2} - 2X(0) + 0^{2}$$

$$Z = X^{2}$$

$$p_{Z}(z) = \begin{cases} (2z)/28 & \text{if } z = 1, 4, 9 \\ z/28 & \text{if } z = 0 \end{cases}$$

(c)
$$\begin{split} E[X^2] &= \sum_{z:P(z)>0} p_Z(z)*z \\ &= (1/14)1 + (4/14)4 + (9/14)9 + (0/28)0 \\ &= 1/14 + 16/14 + 81/14 \\ E[X^2] &= 7 \\ Var(X) &= E[X^2] - (E[X])^2 \\ &= 7 - 0^2 \\ Var(X) &= 7 \end{split}$$

 $Var(x) = \sum_{x=-3}^{3} (x - E[X])^2 p_X(x)$ $= \sum_{x=-3}^{3} (x-0)^2 p_X(x)$

$$= \sum_{x=-3}^{3} x^{2} p_{X}(x)$$

$$= [9(9/28) + 4(4/28) + 1(1/28)] * 2 + 0$$

$$Var(X) = 7$$

Problem 3

(d)

There are a + b - 1 values: $\{2^a, a^{a+1}, \dots, 2^b\}$

$$E[X] = \frac{\sum_{k=a}^{b} 2^k}{b - a + 1}$$

We have a geometric series where r=2, n=a-b+1, $a_1 = 2^a$

$$\begin{split} \sum_{k=a}^{b} 2^k &= 2^a \frac{1 - 2^{b-a+1}}{1 - 2} = 2^{b-1} - 2^a \\ E[X] &= \frac{2^{b+1} - 2^a}{b - a + 1} \end{split}$$

$$E[X] = \frac{2^{b+1} - 2^a}{b - a + 1}$$

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

$$Var(X) = E[X^{2}] - (E[X])^{2}$$

$$E[X^{2}] = \frac{\sum_{k=a}^{b} (2^{2})^{k}}{b-a+1}$$

$$E[X^2] = \frac{4^{b+1} - 4^a}{3(b-a+1)}$$

$$Var(X) = \frac{4^{b+1} - 4^a}{3(b-a+1)} - \left[\frac{2^{b-1} - 2^a}{b-a+1}\right]^2$$

Let X be a random variable where $X \sim Geom(p)$

This is the distribution of candy bars eaten till the first success (getting the golden ticket)

$$f_X = pq^{x-1}$$

$$E[X] = \sum_{x=1}^{\infty} x * pq^{x-1}$$

$$= \sum_{x=1}^{\infty} p \frac{d}{dq} (q^x)$$

$$= p \frac{d}{dq} \sum_{x=1}^{\infty} q^x \quad \text{since the series converges uniformly}$$

$$= p \frac{d}{dq} (\frac{1}{1-q} - 1)$$

$$= p \frac{1}{(1-q)^2} = \frac{1}{p}$$

$$E[X] = \frac{1}{p}$$

$$E[X^{2}] = \sum_{x=1}^{\infty} x^{2} q^{x-1} p$$

$$= \sum_{x=1}^{\infty} ((x-1)+1)^{2} q^{x-1} p$$
since $((x-1)+1)^{2} = (x-1)^{2} + 2(x-1) + 1$ we have:
$$= \sum_{x=1}^{\infty} (x-1)^{2} q^{x-1} p + \sum_{x=1}^{\infty} 2(x-1) q^{x-1} p + \sum_{x=1}^{\infty} q^{x-1} p$$

$$= \sum_{x=1}^{\infty} (x-1)^{2} q^{x-1} p + \left[2 \sum_{x=1}^{\infty} (x-1) q^{x-1} p\right] + 1$$

$$\det i = x-1$$

$$= \sum_{i=0}^{\infty} i^{2} q^{i} p + \left[2 \sum_{i=0}^{\infty} i q^{i} p\right] + 1$$

since i is just a letter we can replace it with x

$$= \left[\sum_{x=1}^{\infty} x^2 q^x p + 0\right] + \left[2\sum_{x=1}^{\infty} x q^x p + 0\right] + 1$$

$$= q \sum_{x=1}^{\infty} x^2 q^{x-1} p + \left[2q \sum_{x=1}^{\infty} x q^{x-1} p\right] + 1$$

$$= q E[X^2] + 2q E[X] + 1$$
Since $E[X] = \frac{1}{p}$

$$E[X^2] = q E[X^2] + (2q) \frac{1}{p} + 1$$

$$(1 - q) E[X^2] = \frac{2q}{p} + 1$$

$$E[X^2] = \frac{2q + p}{p^2} = \frac{q + 1}{p^2}$$

$$Var(X) = \frac{q + 1}{p^2} - \frac{1}{p^2}$$

$$Var(X) = \frac{q}{p^2} = \frac{1 - p}{p^2}$$

Let X be a random variable where $X \sim Geom(1/2)$ This is the distribution of coin tosses until the first success (getting a tails) $P\{X=n\}=p_X(n)=(1-p)^{n-1}p=p^n=(1/2)^n$ Payout= $Y=2^n$ $P\{Y=n\}=P\{X=n\}$ Thus Y has the same PMF as X E[Y] is the expected payout $E[Y]=\sum_{n=1}^{\infty}y\ p_Y(n)$ $E[Y]=\sum_{n=1}^{\infty}2^n(1/2)^n=\sum_{n=1}^{\infty}1=\infty$ Even though the expected payout is infinite there is a very good chance of

Even though the expected payout is infinite there is a very good chance of losing all of your money. Because of this I would not be willing to risk very much. I would be willing to pay \$4 to play since the expected number of tosses is 2.

(a)

$$Y = X^{2} \quad X = \sqrt{Y}$$

$$f_{X}(x) = 1 \quad \text{for } 0 < x < 1$$

$$f_{Y}(y) = f_{x}(g^{-1}(y)) \left| \frac{\mathrm{d}}{\mathrm{d}y}(g^{-1}(y)) \right| \quad y \in \mathcal{Y}$$

$$f_{Y}(y) = (1)|(1/2)y^{-1/2}|$$

$$f_{Y}(y) = \frac{1}{2\sqrt{y}} \quad 0 < y < 1$$

$$f_X(x) = \frac{(n+m+1)!}{n!m!} x^n (1-x)^m \quad 0 < x < 1$$

$$Y = g(X) = -\log(X)$$

$$x = e^{-y} \quad g^{-1}(y) = e^{-y}$$

$$-\log(0) = \infty \quad -\log(1) = 0$$

$$0 < y < \infty$$

$$f_Y(y) = \begin{cases} f_X(g^{-1}(y)) \left| \frac{\mathrm{d}}{\mathrm{d}y}(g^{-1}(y)) \right| & y \in \mathcal{Y} \\ 0 & \text{otherwise} \end{cases}$$

$$f_Y(y) = \begin{cases} f_X(e^{-y}) \left| \frac{\mathrm{d}}{\mathrm{d}y}(e^{-y}) \right| & 0 < y < \infty \\ 0 & \text{otherwise} \end{cases}$$

$$f_Y(y) = \frac{(n+m+1)!}{n!m!} e^{-ny} (1-e^{-y})^m (e^{-y})$$

$$f_Y(y) = \frac{(n+m+1)!}{n!m!} e^{-y(n+1)} (1-e^{-y})^m \quad 0 < y < \infty$$

(c)
$$Y = g(X) = e^{x}$$

$$x = \log(y) \quad g^{-1}(y) = \log(y)$$

$$f_{X}(x) = \frac{1}{\sigma^{2}} x e^{-(x/\sigma)^{2}/2} \quad 0 < x < \infty$$

$$e^{0} = 1 \quad e^{\infty} = \infty$$

$$1 < y < \infty$$

$$f_{Y}(y) = \begin{cases} f_{X}(\log(y)) \left| \frac{d}{dy}(\log(y)) \right| & 1 < y < \infty \\ 0 & \text{otherwise} \end{cases}$$

$$f_{Y}(y) = \frac{1}{\sigma^{2}} \log(y) e^{-(\log(y)/\sigma)^{2}/2} (1/y) \quad 1 < y < \infty$$

(a) Using theorem 2.1.8 we have:

$$A_{0}, A_{1}, A_{2} \text{ of } \chi$$

$$A_{0} = \{0\} \quad A_{1} = (-\infty, 0) \quad A_{2} = (0, \infty)$$

$$P(X \in A_{0}) = 0$$

$$g_{1}(x) = |x|^{3} = -x^{3} \text{ on } A_{1} \quad g_{2}(x) = |x|^{3} = x^{3} \text{ on } A_{2}$$

$$f_{X}(x) = (1/2)e^{-|x|} \quad -\infty < x < \infty$$

$$Y = g_{i}(X)$$

$$g_{1}^{-1}(X) = -Y^{1/3}$$

$$g_{2}^{-1}(X) = Y^{1/3}$$

$$f_{Y}(y) = \sum_{i=1}^{k} f_{X}(g_{i}^{-1}(y)) \left| \frac{d}{dy}(g_{i}^{-1}(y)) \right| \quad y \in \mathcal{Y}$$

$$f_{Y}(y) = 2 * f_{X}(y^{1/3}) \left| \frac{d}{dy}(y^{1/3}) \right| \quad 0 < y < \infty$$

$$f_{Y}(y) = 2 * (1/2)e^{-y^{1/3}}((1/3)y^{-2/3}) \quad 0 < y < \infty$$

$$f_{Y}(y) = (1/3)e^{-y^{1/3}}(y^{-2/3}) \quad 0 < y < \infty$$

$$\int_{0}^{\infty} (1/3)e^{-y^{1/3}}(y^{-2/3}) dy = \int_{0}^{\infty} e^{-y^{1/3}} = 0 + 1 = 1$$

$$A_0 = \{0\} \quad A_1 = (-1,0) \quad A_2 = (0,1)$$

$$P(X \in A_0) = 0$$

$$Y = g_i(X)$$

$$g_1(x) = 1 - X^2 \text{ on } A_1$$

$$g_1^{-1}(x) = -\sqrt{1 - Y}$$

$$g_2(x) = 1 - X^2 \text{ on } A_2$$

$$g_2^{-1}(x) = \sqrt{1 - Y}$$

$$f_X(x) = (3/8)(x+1)^2 - 1 < x < 1$$

$$f_Y(y) = \sum_{i=1}^k f_X(g_i^{-1}(y)) \left| \frac{\mathrm{d}}{\mathrm{d}y}(g_i^{-1}(y)) \right| \quad y \in \mathcal{Y}$$

$$f_Y(y) = (3/8)(\sqrt{1 - y} + 1)^2(1/2)(1 - y)^{-1/2}$$

$$+ (3/8)(-\sqrt{1 - y} + 1)^2(1/2)(1 - y)^{-1/2}$$

$$f_Y(y) = (3/8)[(1 - y)^{1/2} + (1 - y)^{-1/2}] \quad 0 < y < 1$$

$$\int_0^1 (3/8)[(1 - y)^{1/2} + (1 - y)^{-1/2}] dy = \int_0^1 (1/4)(1 - y)^{1/2}(y - 4) = 0 - (1/4 * -4) = 1$$

(c)

$$A_0, A_1, A_2 \text{ of } \chi$$

$$A_0 = \{0\} \quad A_1 = (-1, 0) \quad A_2 = (0, 1)$$

$$f_X(x) = (3/8)(x+1)^2 \quad -1 < x < 1$$

$$Y = 1 - X^2 \text{ if } X \le 0 \quad Y = 1 - X \text{ if } X > 0$$

$$Y = g_i(X)$$

$$g_1(x) = 1 - X^2 \text{ on } A_1$$

$$g_1^{-1}(x) = -\sqrt{1 - Y}$$

$$g_2(x) = 1 - X \text{ on } A_2$$

$$g_2^{-1}(x) = 1 - Y$$

$$f_Y(y) = \sum_{i=1}^k f_X(g_i^{-1}(y)) \left| \frac{\mathrm{d}}{\mathrm{d}y}(g_i^{-1}(y)) \right| \quad y \in \mathcal{Y}$$

$$f_Y(y) = (3/8)(1 - y + 1)^2(1)$$

$$+ (3/8)(-\sqrt{1 - y} + 1)^2(1/2)(1 - y)^{-1/2}$$

$$f_Y(y) = (3/8)(2 - y)^2 + (3/16)(1 - (1 - y)^{1/2})^2(1 - y)^{-1/2} \quad 0 < y < 1$$

$$\int_0^1 (3/8)(2 - y)^2 + (3/16)(1 - (1 - y)^{1/2})^2(1 - y)^{-1/2} dy$$

$$= (3/16) \int_0^1 6 - 8y + 2y^2 + (1 - y)^{1/2} + (1 - y)^{-1/2} dy$$

$$= (3/16) \Big|_0^1 6y - 4y^2 + (2/3)y^3 - (2/3)(1 - y)^{3/2} - (1 - y)^{1/2} = 1$$

Problem 8

Define $Y = F_X(X)$, which means Y is uniformly distributed on (0,1)

$$f(x) = \begin{cases} \frac{x-1}{2} & 1 < x < 3\\ 0 & \text{otherwise} \end{cases}$$

$$\int_{1}^{x} \frac{x-1}{2} = (1/4)(x-1)^{2}$$

$$u(x) = F_{X}(x) = \begin{cases} 0 & x \le 1\\ (1/4)(x-1)^{2} & 1 < x < 3\\ 1 & x \ge 3 \end{cases}$$

Problem 9

(a)

(a)

$$\int_0^\infty (1 - F_X(x)) dx = \int_0^\infty P(X > x) dx$$
Since $P(X > x) = \int_x^\infty f_X(y) dy$

$$\int_0^\infty P(X > x) dx = \int_0^\infty \int_x^\infty f_X(y) dy dx$$

$$= \int_0^\infty \int_0^y dx f_X(y) dy$$

$$= \int_0^\infty (y - 0) f_X(y) dy$$

$$= \int_0^\infty y f_X(y) dy$$

$$= E[X]$$

Since
$$1 - F_X(k) = \sum_{k=1}^{\infty} f_X(k)$$

Then $\sum_{k=0}^{\infty} (1 - F_X(k)) = \sum_{k=0}^{\infty} \left[\sum_{k=1}^{\infty} f_X(k) \right]$
 $= \left(f_X(1) + f_X(2) + \dots \right) + \left(f_X(2) + f_X(3) + \dots \right) + \left(\dots \right)$
 $= 0 + 1 f_X(1) + 2 f_X(2) + 3 f_X(3) + \dots$
 $= \sum_{k=0}^{\infty} k f_X(k) = E[X]$

(a)

$$\int_0^m 3x^2 = \int_m^1 3x^2 = 1/2$$
$$\begin{vmatrix} m & 3 & 1/2 \\ 0 & 1/2 \end{vmatrix} = \frac{1}{m}x^3 = 1/2$$
$$m^3 = 1 - m^3 = 1/2$$
$$m = 2^{-1/3} = 0.7937005$$

Problems 11-14

$$\int_{-\infty}^{m} \frac{1}{\pi(1+x^2)} = \int_{m}^{\infty} \frac{1}{\pi(1+x^2)} = 1/2$$

$$(1/\pi)\Big|_{-\infty}^{m} \arctan(x) = (1/\pi)\Big|_{m}^{\infty} \arctan(x) = 1/2$$

$$(1/\pi)(\arctan(m) + \pi/2) = (1/\pi)(\pi/2 - \arctan(m)) = 1/2$$

$$\arctan(m) = -\arctan(m) = 0$$

$$m = \tan(0) = -\tan(0) = 0$$

$$m = 0$$

WTS: If X is a continuous random variable: $\min_{a} E|X - a| = E|X - m|$

We know
$$E[g(X)] = \int_{-\infty}^{\infty} g(x) f_X(x) dx$$

$$E[X - a] = \int_{-\infty}^{\infty} |x - a| f(x) dx$$

$$= \int_{-\infty}^{a} -(x - a) f(x) dx + \int_{a}^{\infty} (x - a) f(x) dx$$

Setting the first derivative equal to 0 we have:

$$\frac{d}{da}E|X-a|=\int_{-\infty}^af(x)\mathrm{d}x-\int_a^\infty f(x)\mathrm{d}x=0$$
 The solution is:
$$\int_{-\infty}^af(x)\mathrm{d}x=\int_a^\infty f(x)\mathrm{d}x=1/2$$

Which means a=m is the median by definition taking the second derivative we have:

$$\frac{d}{da} \left[\int_{-\infty}^{a} f(x) dx - \int_{a}^{\infty} f(x) dx \right]$$

$$= \frac{d}{da} \left[\Big|_{-\infty}^{a} F(X) - \Big|_{a}^{\infty} F(X) \right]$$

$$= \frac{d}{da} \left[F(a) - F(-\infty) - F(\infty) + F(a) \right]$$

$$= 2f(a) > 0$$

Thus E|X-m| is a minimum by the second derivative test Therefore $\min_a E|X-a|=E|X-m|$

Problem 13

(a) The uniform distribution pdf U(0,1) is symmetric about a=1/2 Standard Normal pdf is symmetric about a=0 The Cauchy distribution pdf is symmetric about a=0

(b)

Let
$$X \sim f(x)$$
 be symmetric at point a That is $\forall \epsilon > 0, f(a + \epsilon) = f(a - \epsilon)$ WTS a is the median:
$$\int_{-\infty}^{a} f(x) \mathrm{d}x = \int_{a}^{\infty} f(x) \mathrm{d}x = 1/2$$
 Proof: Let $\epsilon = x - a$ Then:
$$\int_{a}^{\infty} f(x) \mathrm{d}x = \int_{0}^{\infty} f(a + \epsilon) d\epsilon$$

$$= \int_{0}^{\infty} f(a - \epsilon) d\epsilon = \int_{-\infty}^{a} f(x) \mathrm{d}x \text{ (letting } x = a - \epsilon)$$
 Thus
$$\int_{a}^{\infty} f(x) \mathrm{d}x = \int_{-\infty}^{a} f(x) \mathrm{d}x$$
 We know
$$\int_{-\infty}^{a} f(x) \mathrm{d}x + \int_{a}^{\infty} f(x) \mathrm{d}x = \int_{-\infty}^{\infty} f(x) \mathrm{d}x = 1$$
 Therefore
$$\int_{-\infty}^{a} f(x) \mathrm{d}x = \int_{a}^{\infty} f(x) \mathrm{d}x = 1/2$$

Thus a = median

(c)

Given
$$X \sim f(x)$$
 and symmetric and $E[X]$ exists
Show $E[X] = a$
$$E[X] - a = E[X - a]$$

$$= \int_{-\infty}^{\infty} (x - a) f(x) dx$$

From part b we know a is the median therefore:

$$= \int_{-\infty}^{a} (x - a)f(x)dx + \int_{a}^{\infty} (x - a)f(x)dx$$

making the change of variables $\epsilon = a - x$ for the first integral and $\epsilon = x - a$ for the second.

$$= \int_0^\infty -\epsilon f(a-\epsilon)d\epsilon + \int_0^\infty \epsilon f(a+\epsilon)d\epsilon$$

Since f(x) is symmetric $f(a+e) = f(a-e) \forall \epsilon > 0$

Thus we have:
$$-\int_0^\infty \epsilon f(a-\epsilon)d\epsilon + \int_0^\infty \epsilon f(a-\epsilon)d\epsilon = 0$$

Therefore E[X] - a = E[X - a]

Which means E[X] = a

(d)

Let
$$f(x) = e^{-x}$$
, $x \ge 0$
WTS: For $a > \epsilon > 0$ $f(a + \epsilon) \ne f(a - \epsilon)$
Since $x \ge 0$ we only need to consider positive a $f(a - \epsilon) = e^{-(a - \epsilon)} = e^{-a + \epsilon}$
 $f(a + \epsilon) = e^{-(a + \epsilon)} = e^{-a - \epsilon}$
 $f(a - \epsilon) > f(a + \epsilon)$

Thus f(x) is not symmetric about a > 0

(e)

$$f(x) = e^{-x}, \ x \ge 0$$
 WTS: $median < E[X]$
$$\int_0^m e^{-x} = \int_m^\infty = 1/2$$

$$\Big|_0^m - e^{-x} = \Big|_m^\infty - e^{-x} = 1/2$$

$$-e^{-m} + e^0 = -e^{-\infty} + e^{-m} = 1/2$$

$$-e^{-m} + 1 = e^{-m} = 1/2log(e^{-m}) = log(1/2)$$

$$m = -log(1/2) = log(2)$$

$$f(x) = (1/\lambda)e^{-x/\lambda}, \ \lambda = 1$$

$$E[X] = \int_0^\infty e^{-x/\lambda} dx = \lambda = 1$$

$$log(2) < 1$$

Therefore median < E[X]

Problem 14

- (a) The standard normal distribution is unimodal and has a unique mode, 0.
- (b) The standard uniform distribution U(0,1) is unimodal and does not have a unique mode since every value in the interval is part of the mode.
- (c) Given f(x) is symmetric at point a and unimodal

WTS: a is the mode

Two Cases:

Case 1: Mode=b is unique

Let a be the point of symmetry, b = mode

Assume $a \neq b$, $a = b + \epsilon$, $\epsilon > 0$ that is a is not the mode

By definition of a mode, $f(b) > f(b + \epsilon) \ge f(b + 2\epsilon)$

Which means $f(a - \epsilon) > f(a) \ge f(a + \epsilon)$

This contradicts f(x) being symmetric

Therefore a is the mode.

Case 2: Mode is not unique, b is a mode

That is, there exists an interval (x_1, x_2) where f(x) has the same value in the whole interval.

Assume $a \notin (x_1, x_2)$, $a = (b + \epsilon), \epsilon > 0$, that is a is not a mode

Since b is a mode, $f(b) > f(b + \epsilon) \ge f(b + 2\epsilon)$

Which means $f(a) - \epsilon > f(a) \ge f(a + \epsilon)$ Which contradicts f(x) being symmetric.

Thus a is a mode.

(d) Given $f(x) = e^{-x}, x \ge 0$

WTS: f(x) is unimodal

f(x) is decreasing for $x \geq 0$

Thus for $0 \le x_1 \le x_2$, $f(0) \ge f(x_1) \ge f(x_2)$

Therefore f(x) is unimodal with mode a = 0