## ECON 709 - PS 1

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1. Suppose that  $Y = X^3$  and  $f_X(x) = 42x^5(1-x), x \in (0,1)$ . Find the PDF of Y, and show that the PDF integrates to 1.

Notice that  $Y = X^3$  is a monotone transformation, so we can use the following theorem from the lecture notes:

$$f_Y(y) = \begin{cases} f_X(g^{-1}(y)) | \frac{d}{dy} g^{-1}(y) |, y \in Y \\ 0, \text{ otherwise} \end{cases}$$

$$= \begin{cases} 42(y^{1/3})^5 (1 - y^{1/3}) | (1/3) y^{-2/3} |, y \in (0, 1) \\ 0, \text{ otherwise} \end{cases}$$

$$= \begin{cases} 14y(1 - y^{1/3}), y \in (0, 1) \\ 0, \text{ otherwise} \end{cases}$$

where  $g^{-1}(y) = y^{1/3}$  and  $Y = \{0^3, 1^3\} = \{0, 1\}.$ 

 $f_Y(y)$  integrates to 1:

$$\int_0^1 14t(1-t^{1/3})dt = 14\left[y^2/2 - \frac{y^{7/3}}{7/3}\right]_0^1$$
$$= 14\left[\frac{1}{2} - \frac{3}{7}\right]$$
$$= 1$$

<sup>\*</sup>I worked on this problem set with a study group of Michael Nattinger, Andrew Smith, and Ryan Mather. I also discussed problems with Emily Case, Sarah Bass, and Danny Edgel.

2. For the following CDF and PDF, show that  $f_X$  is the density function of  $F_X$  as long as  $a \ge 0$ . That is, show that for all  $x \in [0,1]$ ,  $F_X(x) = \int_0^x f_X(t) dt$ .

$$F_X(x) = \begin{cases} 1.2x, x \in [0, 0.5) \\ 0.2 + 0.8x, x \in [0.5, 1] \end{cases}$$
$$f_X(x) = \begin{cases} 1.2, x \in [0, 0.5) \\ a, x = 0.5 \\ 0.8, x \in (0.5, 1] \end{cases}$$

Case 1: x < 0.5

$$\int_0^x f_X(t)dt = \int_0^x 1.2dt$$
$$= 1.2x$$
$$= F_X(x)$$

Case 2: x = 0.5

$$\int_0^x f_X(t)dt = \int_0^{0.5} 1.2dt + \int_{0.5}^{0.5} adt$$

$$= 1.2(0.5) + 0$$

$$= 0.6$$

$$= 0.2 + 0.8(0.5)$$

$$= F_X(0.5)$$

Case 3: x > 0.5

$$\int_0^x f_X(t)dt = \int_0^{0.5} 1.2dt + \int_{0.5}^{0.5} adt + \int_{0.5}^x 0.8dt$$
$$= 1.2(0.5) + 0 + 0.8x - 0.8(0.5)$$
$$= 0.6 + 0.8x - 0.4$$
$$= 0.2 + 0.8x$$
$$= F_X(x)$$

3. Let X have PDF  $f_X(x)=\frac{2}{9}(x+1), x\in[-1,2]$ . Find the PDF of  $Y=X^2$ . For  $x\in[-1,2]$ 

$$F_X(x) = \int_{-1}^x \frac{2}{9}(t+1)dt$$

$$= \frac{2}{9} \left[ \frac{t^2}{2} + t \right]_{-1}^x$$

$$= \frac{2}{9} \left[ \frac{x^2}{2} + x - \left( \frac{1}{2} - 1 \right) \right]$$

$$= \frac{x^2}{9} + \frac{2x}{9} + \frac{1}{9}$$

Thus,

$$F_X(x) = \begin{cases} 0, & x < -1\\ \frac{x^2}{9} + \frac{2x}{9} + \frac{1}{9}, & x \in [-1, 2]\\ 1, & x > 2 \end{cases}$$

Consider  $Y = X^2$ . First, notice that  $y \in [0, 4]$ . I consider two cases  $y \in [0, 1]$  and  $y \in (1, 4]$ Case 1:  $y \in [0, 1]$ 

$$F_Y(y) = P(Y \le y)$$

$$= P(X^2 \le y)$$

$$= P(-\sqrt{y} \le X \le \sqrt{y})$$

$$= P(X \le \sqrt{y}) - P(X \le -\sqrt{y})$$

$$= F_X(\sqrt{y}) - F_X(-\sqrt{y})$$

$$= \left[\frac{y}{9} + \frac{2\sqrt{y}}{9} + \frac{1}{9}\right] - \left[\frac{y}{9} - \frac{2\sqrt{y}}{9} + \frac{1}{9}\right]$$

$$= \frac{4\sqrt{y}}{9}$$

Case 2:  $y \in (1, 4]$ 

$$F_Y(y) = P(Y \le y)$$

$$= P(X^2 \le y)$$

$$= P(X \le \sqrt{y})$$

$$= F_X(\sqrt{y})$$

$$= \frac{y}{9} + \frac{2\sqrt{y}}{9} + \frac{1}{9}$$

Thus, the CDF and PDF of Y is:

$$F_X(x) = \begin{cases} 0, & y < 0 \\ \frac{4\sqrt{y}}{9}, & y \in [0, 1] \\ \frac{y}{9} + \frac{2\sqrt{y}}{9} + \frac{1}{9}, & y \in (1, 4] \\ 1, & y > 4 \end{cases}$$

$$f_X(x) = \begin{cases} \frac{2}{0\sqrt{y}}, & y \in [0, 1] \\ \frac{1}{9} + \frac{1}{9\sqrt{y}}, & y \in (1, 4] \\ 0, & \text{otherwise.} \end{cases}$$

4. A median of a distribution is a value m such that  $P(X \le m) \ge 1/2$  and  $P(X \ge m) \ge 1/2$ . Find the median of the distribution  $f(x) = \frac{1}{\pi(1+x^2)}, x \in \mathbb{R}$ .

The CDF of X is

$$F(x) = \int_{-\infty}^{x} \frac{1}{\pi(1+t^2)} dt$$

$$= \frac{1}{\pi} \int_{-\infty}^{x} \frac{1}{1+t^2} dt$$

$$= \frac{1}{\pi} \left[ \tan^{-1}(t) \right]_{-\infty}^{x}$$

$$= \frac{1}{\pi} \left[ \tan^{-1}(x) - \lim_{t \to -\infty} \tan^{-1}(t) \right]$$

$$= \frac{1}{\pi} \left[ \tan^{-1}(x) - \frac{\pi}{2} \right]$$

Now, notice that the distribution is symmetric around 0, so we will consider m=0

$$P(X \le 0) = F(0)$$

$$= \frac{1}{\pi} \left[ \tan^{-1}(0) - \frac{\pi}{2} \right]$$

$$= \frac{1}{\pi} \left[ 0 - \frac{\pi}{2} \right]$$

$$= \frac{1}{2}$$

$$P(X \ge 0) = 1 - P(X \le 0)$$

$$= 1 - F(0)$$

$$= 1 - \frac{1}{2}$$

$$= \frac{1}{2}$$

Thus, m = 0.

5. Show that if X is a continuous random variable, then  $\min_a E|X-a|=E|x-m|$ , where m is the median of X.

$$\begin{split} E|X-a| &= \int_{-\infty}^{\infty} |t-a|f(t)dt \\ &= \int_{-\infty}^{a} (a-t)f(t)dt + \int_{a}^{\infty} (t-a)f(t)dt \\ &= \int_{-\infty}^{a} af(t)dt - \int_{-\infty}^{a} tf(t)dt + \int_{a}^{\infty} tf(t)dt - \int_{a}^{\infty} af(t)dt \\ &= a\bigg(\int_{-\infty}^{a} f(t)dt - \int_{a}^{\infty} f(t)dt\bigg) - \bigg(\int_{-\infty}^{a} tf(t)dt - \int_{a}^{\infty} tf(t)dt\bigg) \\ &= a\bigg(F(a) - (1 - F(a))\bigg) - \bigg(\int_{-\infty}^{a} tf(t)dt - \int_{a}^{\infty} tf(t)dt\bigg) \\ &= a\bigg(2F(a) - 1\bigg) - \bigg(\int_{-\infty}^{a} tf(t)dt - \int_{a}^{\infty} tf(t)dt\bigg) \end{split}$$

Notice that this expression for E|X-a| is differentiable by a. Consider first the second half:

$$\frac{d}{da}\left(\int_{-\infty}^{a} tf(t)dt - \int_{a}^{\infty} tf(t)dt\right) = af(a) - af(a)$$
 = 0

Then the full expression:

$$\frac{d}{da}E|X - a| = \frac{d}{da}a\left(2F(a) - 1\right) + 0$$
$$= \left(2F(a) - 1\right)$$

Setting the derivative equal to zero:

$$2F(a) - 1 = 0$$
$$F(a) = \frac{1}{2}$$

Thus, a = m where  $P(X \ge m) = P(X \le m) = F(m) = \frac{1}{2}$ .

- 6. Let  $\mu_n$  denote the *n*th central moment of a random variable X. Two quantities of interest in addition to the mean and variance are  $\alpha_3 = \frac{\mu_3}{\mu_2^{3/2}}$  and  $\alpha_4 = \frac{\mu_4}{\mu_2^2}$ . The value  $\alpha_3$  is called the skewness and  $\alpha_4$  is called the kurtosis. The skewness measures the lack of symmetry in the density function. The kurtosis, although harder to interpret, measures the peakedness or flatness of the density function.
- (a) Show that if a density function is symmetric about a point a, then  $\alpha_3 = 0$ .

Proof: Define Y = X - a. Y has a symmetric distribution about zero  $\implies E[Y^3] = E[(-Y)^3] = -E[Y^3] = 0$  and E[Y] = 0. Thus, the 3rd central moment of X is zero:

$$\mu_3 = E[(X - E(X))^3]$$

$$= E[(Y + a - E(Y + a))^3]$$

$$= E[(Y + a - E(Y) - a)^3]$$

$$= E[(Y - E(Y))^3]$$

$$= E[Y^3]$$

$$= 0$$

Therefore, the skewness of X is zero:  $\alpha_3 = \frac{0}{\mu_2^{3/2}} = 0$ .

(b) Calculate  $\alpha_3$  for  $f(x) = \exp(-x), x \ge 0$ , a density function that is skewed to the right.

$$M_X(t) = E[e^{tx}]$$

$$= \int_0^\infty e^{tx} e^{-x} dx$$

$$= \int_0^\infty e^{-x(1-t)} dx$$

$$= \left[ \frac{e^{-x(1-t)}}{1-t} (-1) \right]_0^\infty$$

$$= \frac{0}{1-t} (-1) - \frac{1}{1-t} (-1)$$

$$= (1-t)^{-1}$$

where  $0 \le t < 1$ .

$$\begin{split} M_X^{(1)}(t) &= (-1)(1-t)^{-2}(-1) \\ &= (1-t)^{-2} \\ M_X^{(2)}(t) &= (-2)(1-t)^{-3}(-1) \\ &= 2(1-t)^{-3} \\ M_X^{(3)}(t) &= (-3)2(1-t)^{-4}(-1) \\ &= 6(1-t)^{-4} \end{split}$$

The first, second, and third moments of X are:

$$E[X] = M_X^{(1)}(0) = 1$$
  

$$E[X^2] = M_X^{(2)}(0) = 2$$
  

$$E[X^3] = M_X^{(3)}(0) = 6$$

The second and third central moments of X are:

$$\begin{split} &\mu_2 = E[(X - E(X))^2] \\ &= E(X^2) - E(X)^2 \\ &= 2 - 1 \\ &= 1 \\ &\mu_3 = E[(X - E(X))^3] \\ &= E[(X - E(X))(X^2 - 2XE(X) + E(X)^2)] \\ &= E[X^3 - 2X^2E(X) + XE(X)^2 - E(X)X^2 + 2XE(X)^2 - E(X)^3] \\ &= E(X^3) - 2E(X^2)E(X) + E(X)E(X)^2 - E(X)E(X^2) + 2E(X)E(X)^2 - E(X)^3 \\ &= (6) - 2(2)(1) + (1)(1)^2 - (1)(2) + 2(1)(1)^2 - (1)^3 \\ &= 6 - 4 + 1 - 2 + 2 - 1 \\ &= 2 \end{split}$$

Skewness of X is

$$\alpha_3 = \frac{\mu_3}{\mu_2^{3/2}} = \frac{2}{1^{3/2}} = 2$$

(c) Calculate  $\alpha_4$  for the following density functions and comment on the peakedness of each:

$$f(x) = \frac{1}{\sqrt{2\pi}} \exp(-x^2/2), x \in \mathbb{R}$$
(1)

$$f(x) = 1/2, x \in (-1, 1) \tag{2}$$

$$f(x) = \frac{1}{2}\exp(-|x|), x \in \mathbb{R}$$
(3)