## NANYANG TECHNOLOGICAL UNIVERSITY SPMS/DIVISION OF MATHEMATICAL SCIENCES

2016/17 Semester 1 Tutorial 11 MH2500 Probability and Introduction to Statistics

For the tutorial on 3 November, let us discuss

• Ex. 4.7. 70, 73, 77, 80, 85, 96.

**Ex. 4.7.70** If X and Y are independent, show that E(X|Y=y)=E(X).

[Solution:] For discrete case,

$$\begin{split} E(X|Y=y) &= \sum_x x p_{X|Y}(x|y) \\ &= \sum_x \frac{x P(X=x \text{ and } Y=y)}{P(Y=y)} \\ &= \sum_x \frac{x P(X=x) P(Y=y)}{P(Y=y)} \qquad \text{(since $X$ and $Y$ are independent)} \\ &= \sum_x x P(X=x) = E(X). \end{split}$$

For continuous case,

$$\begin{split} E(X|Y=y) &= \int_{-\infty}^{\infty} x f_{X|Y}(x|y) dx \\ &= \int_{-\infty}^{\infty} \frac{x f_{X,Y}(x,y)}{f_Y(y)} dx \\ &= \int_{-\infty}^{\infty} \frac{x f_X(x) f_Y(y)}{f_Y(y)} dx \qquad \text{(since $X$ and $Y$ are independent)} \\ &= \int_{-\infty}^{\infty} x f_X(x) dx = E(X). \end{split}$$

Ex. 4.7.73. A fair coin is tossed n times, and the number of heads, N, is counted. The coin is then tossed N more times. Find the expected total number of heads generated by this process.

[Solution:] Method 1:

Let N denote the number of heads in the first n tosses and let H denote the number of heads in the N tosses. Then N is binomial with parameters n and p and H|N is also binomial with parameters N and p. Therefore, E(H|N) = Np and the expected number is

$$E(N + H) = E(N) + E(H)$$
  
=  $np + E(E(H|N)) = np + E(Np) = np + pE(N) = np + np^2 = np(1 + p).$ 

Since a fair coin is used, p = 1/2 and so  $E(X) = n\frac{1}{2}\frac{3}{2} = \frac{3n}{4}$ .

Method 2: We compute directly. First note that

$$\sum_{l=0}^{k} {l \binom{k}{l}} p^{l} (1-p)^{k-l} = \sum_{l=1}^{k} {k \choose l} p^{l} (1-p)^{k-l} = kp \sum_{l=0}^{k-1} {k-1 \choose l-1} p^{l-1} (1-p)^{k-l} = kp.$$

Let k represents the number of heads in the first n tosses and let l represent the number of heads in the next k tosses. Then computing directly,

$$\begin{split} E(X) &= \sum_{k=0}^{n} \sum_{l=0}^{k} (k+l) \binom{n}{k} \binom{k}{l} p^{k} (1-p)^{n-k} p^{l} (1-p)^{k-l} \\ &= \sum_{k=0}^{n} k \binom{n}{k} p^{k} (1-p)^{n-k} \sum_{l=0}^{k} \binom{k}{l} p^{l} (1-p)^{k-l} + \sum_{k=0}^{n} \binom{n}{k} p^{k} (1-p)^{n-k} \sum_{l=0}^{k} l \binom{k}{l} p^{l} (1-p)^{k-l} \\ &= \sum_{k=0}^{n} k \binom{n}{k} p^{k} (1-p)^{n-k} + p \sum_{k=0}^{n} \binom{n}{k} p^{k} (1-p)^{n-k} k \\ &= np + np^{2} \\ &= np(1+p). \end{split}$$

Ex. 4.7.77. Let X and Y have the joint density

$$f(x,y) = e^{-y}, \qquad 0 \le x \le y.$$

- a. Find Cov(X,Y) and the correlation of X and Y.
- b. Find E(X|Y=y) and E(Y|X=x).
- c. Find the density functions of the random variables E(X|Y) and E(Y|X).

[Solution:]

a. First, the marginal densities are

$$f_X(x) = \int_x^\infty e^{-y} dy = [-e^{-y}]_x^\infty = e^{-x}, \qquad (x \ge 0)$$

$$f_Y(y) = \int_0^y e^{-y} dx = [xe^{-y}]_0^y = ye^{-y} \qquad (y \ge 0).$$

Next, we compute the expected values.

$$\begin{split} E(X) &= \int_0^\infty x e^{-x} dx \ = \ [-x e^{-x}]_0^\infty + \int_0^\infty e^{-x} dx \ = \ 0 + [-e^{-x}]_0^\infty = 1 \\ E(Y) &= \int_0^\infty y^2 e^{-y} dy \ = \ [-y^2 e^{-y}]_0^\infty + \int_0^\infty 2y e^{-y} dy \ = \ 2. \end{split}$$

Similarly, we can show that  $E(X^2)=2$  and  $E(Y^2)=6$  and so  $Var(X)=2-1^2=0$  and  $Var(Y)=6-2^2=2$ .

Hence

$$Cov(X,Y) = \int_0^\infty \int_0^y (x-1)(y-2)e^{-y}dxdy$$

$$= \int_0^\infty \left[ \left( \frac{x^2}{2} - x \right) (y-2)e^{-y} \right]_0^y dy$$

$$= \frac{1}{2} \int_0^\infty (y^3 - 4y^2 + 4y)e^{-y} dy$$

$$= \frac{1}{2} \left\{ [-(y^3 - 4y^2 + 4y)e^{-y}]_0^\infty + \int_0^\infty (3y^2 - 8y + 4)e^{-y} dy \right\}$$

$$= \frac{1}{2} \left\{ 0 + [-(3y^2 - 8y + 4)e^{-y}]_0^\infty + \int_0^\infty (6y - 8)e^{-y} dy \right\}$$

$$= \frac{1}{2} \left\{ 0 + 4 + [-(6y - 8)e^{-y}]_0^\infty + \int_0^\infty 6e^{-y} dy \right\}$$

$$= \frac{1}{2} \left\{ 0 + 4 - 8 + 6[-e^{-y}]_0^\infty \right\} = 1$$

and so the correlation

$$\rho = \frac{\operatorname{Cov}(X, Y)}{\sqrt{\operatorname{Var}(X)}\sqrt{\operatorname{Var}(Y)}} = \frac{1}{\sqrt{2}}.$$

b.

$$E(X|Y=y) = \int_0^y x \frac{e^{-y}}{ye^{-y}} dx = \int_0^y \frac{x}{y} dx = \left[\frac{x^2}{2y}\right]_0^y = \frac{y}{2},$$

$$E(Y|X=x) = \int_x^\infty y \frac{e^{-y}}{e^{-x}} dy = [-ye^{x-y}]_x^\infty + \int_x^\infty e^{x-y} dy = x + [-e^{x-y}]_x^\infty$$

$$= x + 1$$

c. Let U = E(X|Y) and V = E(Y|X). Then by Part b,

$$U = E(X|Y) = \frac{Y}{2}$$
$$V = E(Y|X) = X + 1.$$

Let U = g(Y) = Y/2. Then  $g^{-1}(u) = 2u$  and the density function of U is

$$f_U(u) = f_Y(g^{-1}(u)) \left| \frac{d}{du} g^{-1}(u) \right| = 4ue^{-2u} \qquad (0 \le u < \infty),$$

and  $f_U(u) = 0$  elsewhere. Similarly, let h(X) = X + 1. Then  $h^{-1}(v) = v - 1$  and so

$$f_V(v) = f_X(h^{-1}(v)) \left| \frac{d}{dv} g^{-1}(v) \right| = e^{1-v} \qquad (1 \le v < \infty),$$

and  $f_V(v) = 0$  elsewhere.

Ex. 4.7.80. Let X be a continuous random variable with density function

$$f(x) = 2x, \qquad 0 \le x \le 1.$$

Find the moment-generating function of X, M(t), and verify that E(X) = M'(0) and that  $E(X^2) = M''(0)$ .

[Solution:] The moment generating function is, integrating by parts,

$$M(t) = \int_0^1 e^{tx} 2x \ dx$$

$$= 2 \left[ \frac{1}{t} e^{tx} x \right]_0^1 - 2 \int_0^1 \frac{1}{t} e^{tx} \ dx$$

$$= \frac{2}{t} e^t - \frac{2}{t^2} \left[ e^{tx} \right]_0^1$$

$$= \frac{2}{t} e^t + \frac{2}{t^2} \left( 1 - e^t \right).$$

Therefore,

$$\begin{split} M'(t) &= -\frac{2}{t^2}e^t + \frac{2}{t}e^t - \frac{4}{t^3}(1 - e^t) - \frac{2}{t^2}e^t \\ &= \frac{2}{t}e^t - \frac{4}{t^3}(1 - e^t) - \frac{4}{t^2}e^t \\ &= \frac{2}{t^3}(t^2e^t - 2te^t - 2 + 2e^t) \\ M''(t) &= \frac{2}{t^3}(2te^t + t^2e^t - 2e^t - 2te^t + 2e^t) - \frac{6}{t^4}(t^2e^t - 2te^t - 2 + 2e^t) \\ &= \frac{2}{t^4}(t^3e^t - 3t^2e^t + 6te^t + 6 - 6e^t) \end{split}$$

Therefore, by applying L'Hospital rule once each,

$$\begin{split} M'(0) &= \lim_{t \to 0} \frac{2}{t^3} (t^2 e^t - 2t e^t - 2 + 2e^t) \\ &= \lim_{t \to 0} \frac{2(t^2 e^t + 2t e^t - 2t e^t - 2e^t + 2e^t)}{3t^2} \\ &= \lim_{t \to 0} \frac{2(e^t)}{3} = \frac{2}{3}, \\ M''(0) &= \lim_{t \to 0} \frac{2}{t^4} (t^3 e^t - 3t^2 e^t + 6t e^t + 6 - 6e^t) \\ &= \lim_{t \to 0} \frac{2(3t^2 e^t + t^3 e^t - 6t e^t - 3t^2 e^t + 6e^t + 6t e^t - 6e^t)}{4t^3} \\ &= \lim_{t \to 0} \frac{2t^3 e^t}{4t^3} = \frac{1}{2}. \end{split}$$

Evaluating directly, we find that

$$E(X) = \int_0^1 2x^2 dx = \left[\frac{2}{3}x^3\right]_0^1 = \frac{2}{3}$$

and

$$E(X^2) = \int_0^1 2x^3 dx = \left[\frac{2}{4}x^4\right]_0^1 = \frac{1}{2}.$$

Hence M'(0) = E(X) and  $M''(0) = E(X^2)$ .

Ex. 4.7.85. Find the mgf of a geometric random variable, and use it to find the mean and the variance.

[Solution:]

$$M(t) = \sum_{k=1}^{\infty} e^{tk} (1-p)^{k-1} p = p e^{t} \sum_{k=1}^{\infty} [e^{t} (1-p)]^{k} = \frac{p e^{t}}{1 - e^{t} (1-p)}.$$

Note that in the last equality, we require  $|e^t(1-p)| < 1$  so that the geometric series is convergent. Hence M(t) is only valid for  $e^t < 1/(1-p)$ , i.e.,  $t < \ln(1/(1-p))$ .

By applying the quotient rule, we find that

$$M'(t) = \frac{[1 - e^{t}(1 - p)]pe^{t} + pe^{t}e^{t}(1 - p)}{[1 - e^{t}(1 - p)]^{2}}$$

$$= \frac{pe^{t}}{[1 - e^{t}(1 - p)]^{2}},$$

$$M''(t) = \frac{[1 - e^{t}(1 - p)]^{2}pe^{t} - 2pe^{t}[1 - e^{t}(1 - p)](-e^{t}(1 - p))}{[1 - e^{t}(1 - p)]^{4}}$$

$$= \frac{[1 + e^{t}(1 - p)]pe^{t}}{[1 - e^{t}(1 - p)]^{3}}$$

Setting t = 0 gives

$$M'(0) = \frac{p}{[1 - (1 - p)]^2} = \frac{1}{p},$$
  
$$M''(0) = \frac{[1 + (1 - p)]p}{[1 - (1 - p)]^3} = \frac{2 - p}{p^2}.$$

Hence

$$E(X) = \frac{1}{p}$$
 and  $Var(X) = \frac{2-p}{p^2} - \frac{1}{p^2} = \frac{1-p}{p^2}$ .

**Ex. 4.7.96.** If X and Y have a joint distribution, their joint moment-generating function is defined as

$$M_{XY}(s,t) = E(e^{sX+tY}),$$

which is a function of two variables, s and t.

Show how to find E(XY) from the joint moment-generating function of X and Y.

[Solution:]

The partial derivatives with respect to s and then t is

$$\frac{\partial}{\partial t} \frac{\partial}{\partial s} M_{XY}(s,t) = E(XYe^{sX+tY}).$$

Setting s = t = 0 in the above gives E(XY).