

## Solution of Problem Set 2

### Problem 1:

1.

$$P_X(k) = \binom{n}{k} p^k (1-p)^{n-k}, \quad k = 0, 1, 2, \dots, n$$

$$M_X(s) = \sum_{k=0}^n \binom{n}{k} p^k (1-p)^{n-k} e^{sk} = (1-p + pe^s)^n$$

$$\frac{d}{ds} M_X(s) = npe^s (1-p + pe^s)^{n-1}$$

$$\frac{d^2}{ds^2} M_X(s) = npe^s (1-p + pe^s)^{n-1} + n(n-1)p^2 e^{2s} (1-p + pe^s)^{n-2}$$

$$E(X) = M_X'(0) = np$$

$$E(X^2) = M_X''(0) = np + n(n-1)p^2 = n^2 p^2 + np(1-p)$$

$$\text{Var}(X) = E(X^2) - [E(X)]^2 = np(1-p)$$

2.

$$f_X(x) = \lambda e^{-\lambda x} u(x), \quad k = 1, 2, \dots, n$$

$$M_X(s) = \int_0^{\infty} \lambda e^{-\lambda x} e^{sx} = \frac{\lambda}{\lambda - s}, \quad \text{Re}\{s\} < \lambda$$

$$\frac{d}{ds} M_X(s) = \frac{\lambda}{(\lambda - s)^2}, \quad \text{Re}\{s\} < \lambda$$

$$\frac{d^2}{ds^2} M_X(s) = \frac{2\lambda}{(\lambda - s)^3}, \quad \text{Re}\{s\} < \lambda$$

$$E(X) = M_X'(0) = \frac{1}{\lambda}$$

$$E(X^2) = M_X''(0) = \frac{2}{\lambda^2}$$

$$\text{Var}(X) = E(X^2) - [E(X)]^2 = \frac{2}{\lambda^2} - \frac{1}{\lambda^2} = \frac{1}{\lambda^2}$$

### Problem 2:

For a strictly increasing and differentiable (except for a finite number of points) function:

$$E(Y) = \int_{-\infty}^{\infty} y f_Y(y) dy = \int_{-\infty}^{\infty} y \left( \frac{d}{dy} F_Y(y) \right) dy \stackrel{(1)}{=} \int_{-\infty}^{\infty} y \frac{d}{dy} F_X(g^{-1}(y)) dy =$$

$$\int_{-\infty}^{\infty} y \frac{d}{dy} \left( \int_{-\infty}^{x=g^{-1}(y)} f_X(t) dt \right) dy \stackrel{(2)}{=} \int_{-\infty}^{\infty} y \left( f_X(g^{-1}(y)) \frac{d}{dy} (g^{-1}(y)) \right) dy \stackrel{x=g^{-1}(y)}{=} \int_{-\infty}^{\infty} g(x) f_X(x) dx$$

$dx = \left( \frac{d}{dy} (g^{-1}(y)) \right) dy$

Further explanation:

$$(1) F_Y(y) = P(Y \leq y) = P(g(X) \leq y) = P(X \leq g^{-1}(y)) = F_X(g^{-1}(y))$$

The 3<sup>rd</sup> equality follows from the fact that  $g$  is a strictly increasing function.

(2) The fundamental theorem of calculus claims:

$$V(x) = \int_a^x v(t) dt \quad \Rightarrow \quad \frac{d}{dx} V(x) = v(x)$$

And thus:

$$V(x) = \int_a^{u(x)} v(t) dt \quad \Rightarrow \quad \frac{\partial}{\partial x} V(x) = v(u(x)) \underbrace{\frac{\partial}{\partial x} u(x)}_{\text{נגזרת פנימית}}$$

### **Problem 3:**

The moment-generating function of  $Y = \sum_{i=1}^n X_i$  is

$$M_Y(s) = E(e^{sY}) = E(e^{s(X_1+X_2+\dots+X_n)}) = E\left(\prod_{i=1}^n e^{sX_i}\right) = \prod_{i=1}^n E(e^{sX_i}) = (M_X(s))^n$$

where we swapped the expected value of a product by a product of expected values since  $X_i$  are statistically independent, and the last equality is true due to  $X_i$  being identically distributed hence having the same moment generating function.

### **Problem 4**

1. a. Let us use the transition formula we saw in class:

$$y = e^x$$

For  $y > 0$ , only one solution exists to the equation  $x_1 = \ln(y)$ , and for  $y \leq 0$ , there are no solutions.

Moreover:

$$\left. \frac{de^x}{dx} \right|_{x=\ln(y)} = y$$

All in all:

$$f_Y(y) = \frac{1}{y\sqrt{2\pi\sigma^2}} \exp\left\{-\frac{(\ln(y)-\mu)^2}{2\sigma^2}\right\} u(y)$$

b. Seeing as how  $Y$  distributes like  $e^X$ ,  $X \sim N(\mu, \sigma^2)$ , it follows that  $Y^r$  distributes like  $e^{rX} = e^{X_1}$ , where  $X_1 = rX$ . But from here we get that  $X_1 \sim N(r\mu, r^2\sigma^2)$ , and, therefore, based on section 1 a:  $Y^r \sim LN(r\mu, r^2\sigma^2)$ .

2.

$$Y_1 \sim LN(0,1)$$

$\Rightarrow$  from section 1 b:

$$Y_1^k \sim LN(0, k^2)$$

Consequently:

$$Y_1^k = e^{X_k}, \quad X_k \sim N(0, k^2)$$

$$X_k = kZ, \quad Z \sim N(0,1)$$

$$Y_1^k = e^{kZ}$$

$$E\{Y_1^k\} = E\{e^{kZ}\} = M_Z(k) = e^{\frac{k^2}{2}}$$

where  $M_Z(a)$  is the moment-generating function of a standard Gaussian random variable:

$$M_Z(a) = e^{\frac{a^2}{2}}$$