

# STATS100B – Introduction to Mathematical Statistics

## Homework 3

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### Exercise 1

(a) Solution:

$$M_X(t) = \exp\left\{\frac{\sigma^2}{2}t^2 + \mu t\right\}$$

$$\begin{aligned}M_{aX+b}(t) &= e^{bt}M_X(at) \\&= e^{bt}\exp\left\{\frac{\sigma^2}{2}a^2t + \mu at\right\} \\&= \exp\left\{\frac{(a\sigma)^2}{2}t^2 + (a\mu + b)t\right\} \\&\therefore aX + b \sim N(a\mu + b, a\sigma)\end{aligned}$$

(b) Solution:

$$\begin{aligned}F_Y(y) &= P(aX + b \leq y) \\&= P\left(X \leq \frac{y-b}{a}\right) \\&= \int_{-\infty}^{\frac{y-b}{a}} \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx\end{aligned}$$

$$\begin{aligned}p_Y(y) &= F'_Y(y) \\&= \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{\left[\frac{y-b}{a} - \mu\right]^2}{2\sigma^2}} \frac{1}{a} \\&= \frac{1}{\sqrt{2\pi}(a\sigma)} e^{-\frac{1}{2(a\sigma)^2}[y-(a\mu+b)]^2} \\&\therefore aX + b \sim N(a\mu + b, a\sigma)\end{aligned}$$

## Exercise 2

Solution:

By definition,

$$\begin{aligned}
 M_{\ln X}(t) &= E(e^{t \ln X}) = E(X^t) \\
 &\because \ln X \sim N(\mu, \sigma) \\
 \therefore M_{\ln X}(t) &= \exp\{\mu t + \frac{1}{2}\sigma^2 t^2\} \\
 \therefore E(X^t) &= \exp\{\mu t + \frac{1}{2}\sigma^2 t^2\} \\
 E(X) &= \exp\{\mu + \frac{1}{2}\sigma^2\}, \quad E(X^2) = \exp\{2\mu + 2\sigma^2\} \\
 \text{var}(X) &= E(X^2) - [E(X)]^2 = e^{\sigma^2 + 2\mu}(e^{\sigma^2} - 1)
 \end{aligned}$$

## Exercise 3

$$\begin{aligned}
 M_X(t) &= (1 - \beta t)^{-\alpha} \\
 M_T(t) &= \prod_{i=1}^n M_{X_i}(t) = (1 - \beta t)^{-n\alpha} \\
 \therefore T &\sim \Gamma(n\alpha, \beta) \\
 M_{\bar{X}}(t) &= \prod_{i=1}^n M_{X_i}\left(\frac{t}{n}\right) = \left(1 - \frac{\beta}{n}t\right)^{-n\alpha} \\
 \therefore T &\sim \Gamma\left(n\alpha, \frac{\beta}{n}\right)
 \end{aligned}$$

## Exercise 4

$$\begin{aligned}
 E(X^4) &= E[X^3(X - \mu + \mu)] \\
 &= E[X^3(X - \mu)] + \mu E(X^3) \\
 &= \sigma^2 E(X^2) + \mu \left[ E[(X^2)(X - \mu)] + \mu E(X^2) \right] \\
 &= \sigma^2 [(EX)^2 + \text{var}(X)] + \mu \left[ 2\sigma^2 E(X) + \mu [(EX)^2 + \text{var}(X)] \right] \\
 &= \sigma^2 [\mu^2 + \sigma^2] + \mu \left[ 2\mu\sigma + \mu [\mu^2 + \sigma^2] \right] \\
 &= \mu^2 + 6\mu^2\sigma^2 + 3\sigma^4
 \end{aligned}$$

### Exercise 5

$$X = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \sim N\left(\begin{pmatrix} \mu_1 \\ \mu_2 \end{pmatrix}, \begin{pmatrix} \sigma_1 & \sigma_{12} \\ \sigma_{21} & \sigma_2 \end{pmatrix}\right)$$

$$M_X(t) = M_{X_1, X_2}(t_1, t_2) = E(e^{t_1 X_1 + t_2 X_2}) = \exp\left\{t_1 \mu_1 + t_2 \mu_2 + \frac{1}{2} \sigma_1^2 t_1^2 + \frac{1}{2} \sigma_2^2 t_2^2 + t_1 t_2 \sigma_{12}\right\}$$

$$\frac{\partial^2 \Psi(t_1, t_2)}{\partial t_1 \partial t_2} = \sigma_{12}$$

$$\therefore \frac{\partial^2 \Psi(0, 0)}{\partial t_1 \partial t_2} = \sigma_{12} = \text{cov}(X_1, X_2)$$

$$\therefore \frac{\partial^2 M_{X_1, X_2}(t_1, t_2)}{\partial t_1 \partial t_2} = \sigma_{12} e^{\Psi(t_1, t_2)} + (\mu_1 + \sigma_1^2 t_1 + t_1 \sigma_{12})(\mu_2 + \sigma_2^2 t_2 + t_1 \sigma_{12})$$

$$\frac{\partial M_{X_1, X_2}(t_1, t_2)}{\partial t_1} = (\mu_1 + \sigma_1^2 t_1 + t_1 \sigma_{12}) e^{\Psi(t_1, t_2)}$$

$$\frac{\partial M_{X_1, X_2}(t_1, t_2)}{\partial t_2} = (\mu_2 + \sigma_2^2 t_2 + t_2 \sigma_{12}) e^{\Psi(t_1, t_2)}$$

$$\Psi(0, 0) = 0$$

$$\begin{aligned} \therefore \frac{\partial^2 M_{X_1, X_2}(0, 0)}{\partial t_1 \partial t_2} - \frac{\partial M_{X_1, X_2}(0, 0)}{\partial t_1} \times \frac{\partial M_{X_1, X_2}(0, 0)}{\partial t_2} &= (\sigma_{12} + \mu_1 \mu_2) - \mu_1 \times \mu_2 \\ &= \sigma_{12} \\ &= \text{cov}(X_1, X_2) \end{aligned}$$