

MATH20802: STATISTICAL METHODS
SECOND SEMESTER
ANSWERS TO THE IN CLASS TEST

1. For a random variable X let $M_X(t) = E[\exp(tX)]$ denote its moment generating function (mgf).

(i) if $X \sim \text{Bin}(n, p)$ with the pmf $p(x) = \binom{n}{x} p^x (1-p)^{n-x}$ for $x = 0, 1, \dots, n$ and $0 < p < 1$ then

$$\begin{aligned} M_X(t) &= \sum_{x=0}^n \exp(tx) \binom{n}{x} p^x (1-p)^{n-x} \\ &= (1-p)^n \sum_{x=0}^n \binom{n}{x} \left\{ \frac{p \exp(t)}{1-p} \right\}^x \\ &= (1-p)^n \left\{ 1 + \frac{p \exp(t)}{1-p} \right\}^n \\ &= \{1 - p + p \exp(t)\}^n. \end{aligned}$$

(ii) if $X \sim \text{Po}(\theta)$ with the pmf $p(x) = \theta^x \exp(-\theta)/x!$ for $x = 0, 1, \dots$ then

$$\begin{aligned} M_X(t) &= \sum_{x=0}^{\infty} \exp(tx) \frac{\theta^x \exp(-\theta)}{x!} \\ &= \exp(-\theta) \sum_{x=0}^{\infty} \frac{1}{x!} (\theta \exp(t))^x \\ &= \exp(-\theta) \exp(\theta \exp(t)) \\ &= \exp\{\theta [\exp(t) - 1]\}. \end{aligned}$$

(iii) if $X \sim \text{Cauchy}(0, 1)$ with the pdf $f(x) = (1/\pi)(1+x^2)^{-1}$ for $-\infty < x < \infty$ then

$$M_X(t) = \frac{1}{\pi} \int_{-\infty}^{\infty} \frac{\exp(tx)}{1+x^2} dx = 1$$

if $t = 0$;

$$\begin{aligned} M_X(t) &= \frac{1}{\pi} \int_{-\infty}^{\infty} \frac{\exp(tx)}{1+x^2} dx \\ &\geq \frac{1}{\pi} \int_0^{\infty} \frac{\exp(tx)}{1+x^2} dx \\ &\geq \frac{t}{2\pi} \int_0^{\infty} \frac{2x}{1+x^2} dx \\ &= \frac{t}{2\pi} \left[\log(1+x^2) \right]_0^{\infty} \\ &= \infty \end{aligned}$$

if $t > 0$; and,

$$M_X(t) = \frac{1}{\pi} \int_{-\infty}^{\infty} \frac{\exp(tx)}{1+x^2} dx$$

$$\begin{aligned}
&\geq \frac{1}{\pi} \int_{-\infty}^0 \frac{\exp(tx)}{1+x^2} dx \\
&\geq \frac{t}{2\pi} \int_{-\infty}^0 \frac{2x}{1+x^2} dx \\
&= \frac{t}{2\pi} [\log(1+x^2)]_{-\infty}^0 \\
&= \infty
\end{aligned}$$

if $t < 0$.

- (iv) if $X \sim \text{Gamma}(a, b)$ with the pdf $f(x) = b^a x^{a-1} \exp(-bx)/\Gamma(a)$ for $x > 0$, $a > 0$ and $b > 0$, where $\Gamma(\cdot)$ denotes the gamma function, then

$$\begin{aligned}
M_X(t) &= \frac{b^a}{\Gamma(a)} \int_0^\infty x^{a-1} \exp\{-(b-t)x\} dx \\
&= \frac{b^a}{(b-t)^a \Gamma(a)} \int_0^\infty y^{a-1} \exp\{-y\} dy \\
&= \frac{b^a}{(b-t)^a \Gamma(a)} \Gamma(a) \\
&= \left(\frac{b}{b-t}\right)^a.
\end{aligned}$$

- (v) if $X \sim \text{Beta}(3, 1)$ with the pdf $f(x) = 3x^2$ for $0 < x < 1$ then

$$\begin{aligned}
M_X(t) &= 3 \int_0^1 x^2 \exp(tx) dx \\
&= 3 \left\{ \left[\frac{x^2 \exp(tx)}{t} \right]_0^1 - \frac{2}{t} \int_0^1 x \exp(tx) dx \right\} \\
&= 3 \left\{ \frac{\exp(t)}{t} - \frac{2}{t} \left\{ \left[\frac{x \exp(tx)}{t} \right]_0^1 - \frac{1}{t} \int_0^1 \exp(tx) dx \right\} \right\} \\
&= 3 \left\{ \frac{\exp(t)}{t} - \frac{2 \exp(t)}{t^2} + \frac{2}{t^2} \int_0^1 \exp(tx) dx \right\} \\
&= 3 \left\{ \frac{\exp(t)}{t} - \frac{2 \exp(t)}{t^2} + \frac{2 [\exp(t) - 1]}{t^3} \right\}.
\end{aligned}$$

2. Suppose $X_1, X_2, \dots, X_n$ is a random sample from $\text{Uni}[a, b]$, where both a and b are unknown.

- (i) The joint likelihood function of a and b is

$$\begin{aligned}
L(a, b) &= \frac{1}{b-a} I\{a < X_1 < b\} \frac{1}{b-a} I\{a < X_2 < b\} \cdots \frac{1}{b-a} I\{a < X_n < b\} \\
&= \frac{1}{(b-a)^n} \prod_{i=1}^n I\{a < X_i < b\} \\
&= \frac{1}{(b-a)^n} I\{\max(X_1, X_2, \dots, X_n) < b\} I\{a < \min(X_1, X_2, \dots, X_n)\},
\end{aligned}$$

where $I\{\cdot\}$ denotes the indicator function.

- (ii) Note that $(b-a)^{-n}$ is an increasing function of a over $(-\infty, \min(X_1, X_2, \dots, X_n))$. So, the maximum of $L(a, b)$ will be attained at $a = \min(X_1, X_2, \dots, X_n)$.

- (iii) Note that $(b-a)^{-n}$ is a decreasing function of b over $(\max(X_1, X_2, \dots, X_n), \infty)$. So, the maximum of $L(a, b)$ will be attained at $b = \max(X_1, X_2, \dots, X_n)$.
- (iv) Let $Z = \min(X_1, X_2, \dots, X_n)$. Note that

$$F_Z(z) = 1 - \left(\frac{b-z}{b-a} \right)^n$$

and

$$f_Z(z) = n \frac{(b-z)^{n-1}}{(b-a)^n}$$

for $a < z < b$. Furthermore,

$$\begin{aligned} E(Z) &= n \int_a^b z \frac{(b-z)^{n-1}}{(b-a)^n} dz \\ &= n \int_a^b (b - (b-z)) \frac{(b-z)^{n-1}}{(b-a)^n} dz \\ &= nb \int_a^b \frac{(b-z)^{n-1}}{(b-a)^n} dz - n \int_a^b \frac{(b-z)^n}{(b-a)^n} dz \\ &= b - \frac{n(b-a)}{n+1} \\ &= \frac{na}{n+1} + \frac{b}{n+1}, \end{aligned}$$

$$\begin{aligned} E(Z^2) &= n \int_a^b z^2 \frac{(b-z)^{n-1}}{(b-a)^n} dz \\ &= n \int_a^b (b - (b-z))^2 \frac{(b-z)^{n-1}}{(b-a)^n} dz \\ &= nb^2 \int_a^b \frac{(b-z)^{n-1}}{(b-a)^n} dz - 2nb \int_a^b \frac{(b-z)^n}{(b-a)^n} dz + n \int_a^b \frac{(b-z)^{n+1}}{(b-a)^n} dz \\ &= b^2 - \frac{2nb(b-a)}{n+1} + \frac{n(b-a)^2}{n+2}, \end{aligned}$$

and

$$Var(Z) = \left[\frac{n}{n+2} - \frac{n^2}{(n+1)^2} \right] (b-a)^2.$$

So, it follows that the mle, $\hat{a}$, is a biased and consistent estimator for a .

- (v) Let $Z = \max(X_1, X_2, \dots, X_n)$. Note that

$$F_Z(z) = \left(\frac{z-a}{b-a} \right)^n$$

and

$$f_Z(z) = n \frac{(z-a)^{n-1}}{(b-a)^n}$$

for $a < z < b$. Furthermore,

$$\begin{aligned}
E(Z) &= n \int_a^b z \frac{(z-a)^{n-1}}{(b-a)^n} dz \\
&= n \int_a^b (z-a+a) \frac{(z-a)^{n-1}}{(b-a)^n} dz \\
&= n \int_a^b \frac{(z-a)^n}{(b-a)^n} dz + na \int_a^b \frac{(z-a)^{n-1}}{(b-a)^n} dz \\
&= \frac{n(b-a)}{n+1} + a \\
&= \frac{a}{n+1} + \frac{nb}{n+1},
\end{aligned}$$

$$\begin{aligned}
E(Z^2) &= n \int_a^b z^2 \frac{(z-a)^{n-1}}{(b-a)^n} dz \\
&= n \int_a^b (z-a+a)^2 \frac{(z-a)^{n-1}}{(b-a)^n} dz \\
&= n \int_a^b \frac{(z-a)^{n+1}}{(b-a)^n} dz + 2na \int_a^b \frac{(z-a)^n}{(b-a)^n} dz + na^2 \int_a^b \frac{(z-a)^{n-1}}{(b-a)^n} dz \\
&= \frac{n(b-a)^2}{n+2} + \frac{2na(b-a)}{n+1} + a^2,
\end{aligned}$$

and

$$Var(Z) = \left[\frac{n}{n+2} - \frac{n^2}{(n+1)^2} \right] (b-a)^2.$$

So, it follows that the mle, $\hat{b}$, is a biased and consistent estimator for b .