

Introduction to mathematical Statistics Lecture 3

***1. Review: The derivative and integral of some important functions one should remember.**

$$\frac{d}{dx}(x^k) = kx^{k-1}$$

$$\frac{d}{dx}(e^x) = e^x$$

$$\frac{d}{dx}(\ln x) = \frac{1}{x}$$

$$\int_a^b x^k dx = \left(\frac{1}{k+1} x^{k+1} \right) \Big|_{x=a}^{x=b} = \frac{1}{k+1} (b^{k+1} - a^{k+1})$$

$$\int_a^b e^x dx = e^x \Big|_{x=a}^{x=b} = e^b - e^a$$

***The Chain Rule**

$$\frac{d}{dx} g[f(x)] = g'[f(x)] \cdot f'(x)$$

For example: $\frac{d}{dx} e^{x^2} = e^{x^2} \cdot 2x$

***The Product Rule**

$$\frac{d}{dx} [g(x) \cdot f(x)] = g'(x)f(x) + g(x)f'(x)$$

***2. MGF, its second function: The m.g.f. will also generate the moments**

Moment: 1st (population) moment $E(X) = \int x \cdot f(x) dx$

2nd (population) moment $E(X^2) = \int x^2 \cdot f(x) dx$

...

Kth (population) moment $E(X^k) = \int x^k \cdot f(x) dx$

Take the Kth derivative of the $M_X(t)$ at $t=0$, we get the Kth moment of X

$$\left. \frac{d}{dt} M_X(t) \right|_{t=0} = E(X)$$

$$\left. \frac{d^2}{dt^2} M_X(t) \right|_{t=0} = E(X^2)$$

...

$$\left. \frac{d^k}{dt^k} M_X(t) \right|_{t=0} = E(X^k)$$

Example: When $X \sim N(\mu, \sigma^2)$, we want to verify the above equations for $k=1$ & $k=2$.

$$\frac{d}{dt} M_X(t) = (e^{\mu + \frac{1}{2}\sigma^2 t^2}) \cdot (\mu + \sigma^2 t) \quad (\text{using the chain rule})$$

So when $t=0$

$$\left. \frac{d}{dt} M_X(t) \right|_{t=0} = \mu = E(X)$$

$$\frac{d^2}{dt^2} M_X(t)$$

$$= \frac{d}{dt} \left[\frac{d}{dt} M_X(t) \right]$$

(using the result of the Product Rule)

$$= \frac{d}{dt} [(e^{\mu + \frac{1}{2}\sigma^2 t^2}) \cdot (\mu + \sigma^2 t)]$$

$$= (e^{\mu + \frac{1}{2}\sigma^2 t^2}) \cdot (\mu + \sigma^2 t)^2 + (e^{\mu + \frac{1}{2}\sigma^2 t^2}) \cdot \sigma^2$$

And

$$\left. \frac{d^2}{dt^2} M_X(t) \right|_{t=0}$$

$$= \mu^2 + \sigma^2 \quad \text{Considering } \sigma^2 = \text{Var}(X) = E(X^2) - \mu^2$$

$$= E(X^2)$$

***3. Review: Bivariate Normal Random Variable**

$(X, Y) \sim BN(\mu_x, \sigma_x^2; \mu_y, \sigma_y^2; \rho)$ where ρ is the correlation between X & Y

The joint p.d.f. of (X, Y) is

$$f(x, y) = \frac{1}{2\pi\sigma_x\sigma_y\sqrt{1-\rho^2}} \exp\left(-\frac{1}{2(1-\rho^2)}\left(\frac{x^2}{\sigma_x^2} + \frac{y^2}{\sigma_y^2} - \frac{2\rho xy}{\sigma_x\sigma_y}\right)\right)$$

4. Homework 2 (Solution)

Q1.

Let X_1 & X_2 be independent $N(\mu, \sigma^2)$

prove that $M_{X_1+X_2, X_1-X_2}(t_1, t_2) = M_{X_1+X_2}(t_1)M_{X_1-X_2}(t_2)$

Solution:

Let $W = X_1 + X_2$ and $Y = X_1 - X_2$

W and Y are independent if and only if $M_{W,Y}(t_1, t_2) = M_W(t_1)M_Y(t_2)$

$$M_{W,Y}(t_1, t_2) = E(e^{t_1 W + t_2 Y}) \xrightarrow{\text{substitute for W \& Y}} E(e^{t_1(X_1 + X_2) + t_2(X_1 - X_2)}) = E(e^{(t_1 + t_2)X_1 + (t_1 - t_2)X_2})$$

$E(e^{(t_1 + t_2)X_1 + (t_1 - t_2)X_2}) = E(e^{(t_1 + t_2)X_1})E(e^{(t_1 - t_2)X_2})$ because X_1 & X_2 are independent

$$\bullet E(e^{(t_1 + t_2)X_1}) = e^{\mu(t_1 + t_2) + \frac{\sigma^2}{2}(t_1 + t_2)^2} = e^{\mu t_1 + \mu t_2 + \frac{\sigma^2}{2}t_1^2 + \sigma^2 t_1 t_2 + \frac{\sigma^2}{2}t_2^2}$$

$$\bullet E(e^{(t_1 - t_2)X_2}) = e^{\mu(t_1 - t_2) + \frac{\sigma^2}{2}(t_1 - t_2)^2} = e^{\mu t_1 - \mu t_2 + \frac{\sigma^2}{2}t_1^2 - \sigma^2 t_1 t_2 + \frac{\sigma^2}{2}t_2^2}$$

$$\begin{aligned}
 \therefore E(e^{(t_1+t_2)X_1+(t_1-t_2)X_2}) &= e^{\mu t_1+\mu t_2+\frac{\sigma^2 t_1^2}{2}+\sigma^2 t_1 t_2+\frac{\sigma^2 t_2^2}{2}} e^{\mu t_1-\mu t_2+\frac{\sigma^2 t_1^2}{2}-\sigma^2 t_1 t_2+\frac{\sigma^2 t_2^2}{2}} \\
 &= \exp\left(\mu t_1+\mu t_2+\frac{\sigma^2 t_1^2}{2}+\sigma^2 t_1 t_2+\frac{\sigma^2 t_2^2}{2}+\mu t_1-\mu t_2+\frac{\sigma^2 t_1^2}{2}-\sigma^2 t_1 t_2+\frac{\sigma^2 t_2^2}{2}\right) \\
 &= \exp\left(2\mu t_1+\frac{1}{2}(2\sigma^2)t_1^2+\frac{1}{2}(2\sigma^2)t_2^2\right) \\
 &= e^{2\mu t_1+\frac{1}{2}(2\sigma^2)t_1^2} e^{\mu t_2+\frac{1}{2}(2\sigma^2)t_2^2} = M_{X_1+X_2}(t_1)M_{X_1-X_2}(t_2) \\
 &= M_{1W}(t_1)M_Y(t_2) \\
 \therefore M_{1W,Y}(t_1,t_2) &= M_{1W}(t_1)M_Y(t_2)
 \end{aligned}$$

$\therefore W = X_1 + X_2$ and $Y = X_1 - X_2$ are independent

Q2. Let Z be $N(0,1)$, (1) please derive the covariance between Z and Z^2 ; (2) Are Z and Z^2

independent?

Solution:

$$(1) \text{cov}(Z, Z^2) = E[(Z - E(Z))(Z^2 - E(Z^2))]$$

$$= E[(Z - 0)(Z^2 - 1)] = E[Z^3] = \frac{d^3}{dt^3} \left[\exp\left(\frac{t^2}{2}\right) \right] \Big|_{t=0} = 1$$

(2) Z and Z^2 are not independent. You can do it in many ways, using (a) pdf, (b)

cdf/probabilities, or (c) mgf.

For example, $P(Z > 1, Z^2 > 1) = P(Z > 1) \neq P(Z > 1) * P(Z^2 > 1) = [P(Z > 1)]^2$

Q3. Let $X \sim N(3,4)$, please calculate $P(1 < X < 3)$

Solution:

$$\begin{aligned} &P(1 < X < 3) \\ &= P\left(\frac{1-3}{2} < \frac{X-3}{2} < \frac{3-3}{2}\right) \\ &= P(-1 < Z < 0) \\ &= P(0 < Z < 1) \\ &= 0.3413 \end{aligned}$$

Note: In the above, $Z \sim N(0,1)$