

# Lecture 19

11/22/17

$$Z \sim N(0,1) = \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}$$

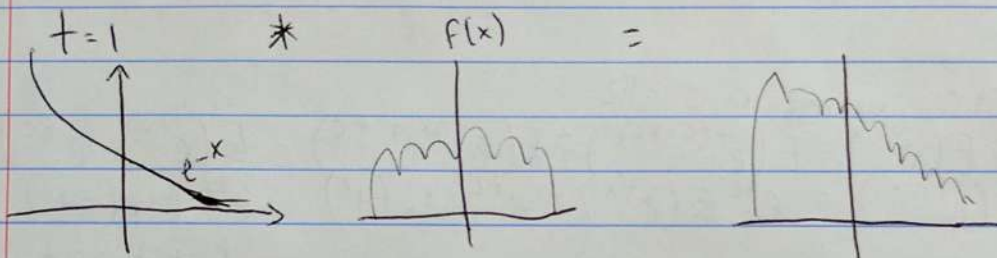
$$X = \mu + \sigma Z \sim N(\mu, \sigma^2) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1}{2\sigma^2}(x-\mu)^2}$$

$$\text{Supp}(Z) = \text{Supp}(X) = \mathbb{R}$$

$$Z = \frac{x-\mu}{\sigma}$$

Z scores

$$L(t) = B(f) = \int_{\mathbb{R}} e^{tx} f(x) dx - \text{Bilateral Laplace Transformation}$$



Thm  $L(t)$  and  $f(x)$  are 1:1 (if  $L(t)$  exists).

$$M_X(t) := E(e^{tx}) = \int_{\mathbb{R}} e^{tx} f(x) dx \text{ if } X \text{ is continuous}$$

$$= \sum_{x \in \text{Supp}(X)} e^{tx} p(x) \text{ if } X \text{ is discrete}$$

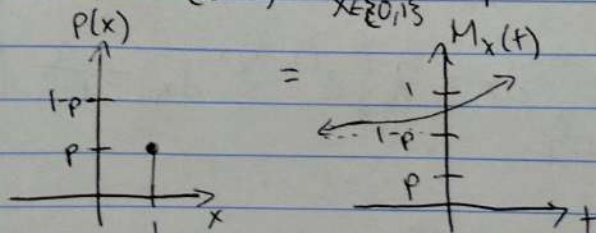
$$\textcircled{1} M_X(t) = M_Y(t) \Leftrightarrow X \stackrel{d}{=} Y$$

$\Leftarrow$  uses properties we know

$\Rightarrow$  theorem: equation of Bilateral Laplace Transformation

$$X \sim \text{Bern}(p)$$

$$M_X(t) = E(e^{tx}) = \sum_{x \in \{0,1\}} e^{tx} p^x (1-p)^{1-x} = 1-p + pe^t$$





$\textcircled{II} M_X(t) = E(e^{tx})$   
 $M'_X(t) = \frac{d}{dt}(E(e^{tx})) = \frac{d}{dt} \left( \int_{\mathbb{R}} e^{tx} f(x) dx \right) = \int_{\mathbb{R}} \frac{d}{dt} (e^{tx} f(x)) dx$  Sometimes yes  
Sometimes no  
 $= \int_{\mathbb{R}} x e^{tx} f(x) dx = E(X e^{tx})$

$M'_X(0) = E(X)$  - First moment

$M''_X(t) = E(X^2 e^{tx})$

$M''_X(0) = E(X^2)$  - Second moment

$M^{(k)}_X(0) = E(X^k)$  -  $k^{\text{th}}$  moment

$\textcircled{III} Y = ax + c$ , where  $a, c \in \mathbb{R}$   
 $M_Y(t) = E(e^{tY}) = E(e^{t(ax+c)}) = E(e^{tax+tc}) = E(e^{tax} \cdot e^{tc})$   
 $= e^{tc} E(e^{tax}) = e^{tc} E(e^{t'x}) = e^{tc} M_X(t') = e^{tc} M_X(at)$   
Let  $t' = at$

$\textcircled{IV} Y = X_1 + X_2$  such that  $X_1, X_2$  are independent  
 $M_Y(t) = E(e^{tY}) = E(e^{t(X_1+X_2)}) = E(e^{tX_1} e^{tX_2}) = E(e^{tX_1}) E(e^{tX_2})$   
 $= M_{X_1}(t) M_{X_2}(t) = (M_X(t))^2$   
if iid

$X \sim \text{Binomial}(n, p)$   
 $M_X(t) = E(e^{tx}) = \sum_{x=0}^n e^{tx} \binom{n}{x} p^x (1-p)^{n-x} = \sum_{x=0}^n \binom{n}{x} (pe^t)^x (1-p)^{n-x}$   
 $= (1-p + pe^t)$

$X_1, \dots, X_n \stackrel{\text{iid}}{\sim} \text{Bern}(p)$

$T = X_1 + \dots + X_n \sim \text{Binomial}$

$M_T(t) = (M_X(t))^n = (1-p + pe^t)^n \Rightarrow T \sim \text{Binomial}(n, p)$   
 $\textcircled{II}$   $\textcircled{I}$

$X \sim \text{Exp}(\lambda)$

$M_X(t) = E(e^{tx}) = \int_0^\infty e^{tx} \lambda e^{-\lambda x} dx = \lambda \int_0^\infty e^{(t-\lambda)x} dx = \frac{\lambda}{t-\lambda} (e^{(t-\lambda)x})_0^\infty$

If  $\lambda > t \Rightarrow \frac{\lambda}{t-\lambda} (0-1) = \frac{\lambda}{\lambda-t}$

$X \sim \text{Exp}(\lambda)$

$Y = aX, a > 0$  (scaling)

$M_Y(t) = M_X(at) = \frac{\lambda}{\lambda-at} \cdot \frac{1/a}{1/a} = \frac{\lambda/a}{\lambda/a - t} = \frac{\lambda'}{\lambda' - t} \Rightarrow Y \sim \text{Exp}(\lambda') = \text{Exp}\left(\frac{\lambda}{a}\right)$   
 $\textcircled{III}$   $\textcircled{II}$

$$Z \sim N(0, 1)$$

$$M_Z(t) = E(e^{tz}) = \int_{\mathbb{R}} e^{tx} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx = \int_{\mathbb{R}} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2} + tx} dx = \int_{\mathbb{R}} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}(x^2 - 2tx)} dx$$

$$ax^2 + bx + c = (x+d)^2 + e \quad (x-t)^2 = x^2 - 2tx + t^2$$

$$ax^2 + bx = (x+d)^2 + e \quad x^2 - 2tx = (x-t)^2 - t^2$$

$$\rightarrow \int_{\mathbb{R}} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}(x-t)^2 - t^2} dx = \int_{\mathbb{R}} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}(x-t)^2} e^{-t^2} dx = e^{-\frac{t^2}{2}} \underbrace{\int_{\mathbb{R}} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}(x-t)^2} dx}_{\text{PDF of } N(t, 1)} = e^{-\frac{t^2}{2}}$$

$$\text{Verify } E(Z) = 0$$

$$M'_Z(0) = t e^{\frac{t^2}{2}} \Big|_0 = 0$$

$$\text{Verify } SE(Z) = 1 \Rightarrow \text{Var}(Z) = 1$$

$$E(Z^2) = 1$$

$$M''_Z(0) = e^{\frac{t^2}{2}} + t^2 e^{\frac{t^2}{2}} \Big|_0 = 1$$