

## warmup

Show that the MGF of  $X \sim \text{Poi}(\mu)$  is  
 $M_X(t) = e^{\mu(e^t - 1)}$  for all  $t$ .

Recall  $M_X(t) = E(e^{tX}) \leftarrow$  think of  $e^{tX}$  as  $g(X)$ ,  
 and  $P(X=k) = \frac{e^{-\mu} \mu^k}{k!}$

$$\begin{aligned} M_X(t) = E(e^{tX}) &= \sum_{k=0}^{\infty} e^{tk} P(X=k) = \sum_{k=0}^{\infty} e^{tk} \frac{e^{-\mu} \mu^k}{k!} \\ &= e^{-\mu} \sum_{k=0}^{\infty} \frac{e^{tk} \mu^k}{k!} = e^{-\mu} \sum_{k=0}^{\infty} \frac{(e^t \mu)^k}{k!} \end{aligned}$$

Recall from Calculus

$$e^a = \sum_{k=0}^{\infty} \frac{a^k}{k!} \quad \text{Taylor for all } a \in \mathbb{R}$$

$$\begin{aligned} &= e^{-\mu} e^{e^t \mu} \\ &= \boxed{e^{\mu(e^t - 1)} \text{ for all } t} \end{aligned}$$

Note

$$E(X) = \left. \frac{d}{dt} M_X(t) \right|_{t=0} = e^{\mu(e^t - 1)} \cdot \mu e^t \Big|_{t=0} = \boxed{\mu}$$

Announcement: Q2

Tuesday October 24

cover >

Sec 4.1, 4.2, 4.4, 4.5, MGF

## Last time

### Key Properties of MGF

- ① If an MGF exists in an interval containing zero,  $M^{(k)}(t) \Big|_{t=0} = E(X^k)$

last time

- ② If  $X$  and  $Y$  are independent RVs,

$$M_{X+Y}(t) = M_X(t) M_Y(t).$$

Proved in MGF HW.

- ③ If  $M_X(t) = M_Y(t)$  for all  $t$  in an

interval around 0 then  $F_X(z) = F_Y(z)$

(i.e.  $X$  and  $Y$  have the same distribution).

Skip proof — we can invert a MGF to get the CDF.

④  $M_{aX}(t) = M_X(at)$

PF/  $M_{aX}(t) = E(e^{atX})$

$$= E(e^{Xat})$$

$$= M_X(at)$$

□

ex A RV  $X$  takes values 1, 2, 3  
with prob  $\frac{1}{2}, \frac{1}{3}, \frac{1}{6}$ .

Find  $M_X(t)$ .

$$\begin{aligned} E(e^{tx}) &= \sum_{x=1}^3 e^{tx} P(x) \\ &= e^{t \cdot 1} \cdot \frac{1}{2} + e^{2t} \cdot \frac{1}{3} + e^{3t} \cdot \frac{1}{6} \\ &\text{for all } t \end{aligned}$$

$\approx E(e^{tx})$

ex If  $M_X(t) = \frac{1}{2}e^{1t} + \frac{1}{3}e^{2t} + \frac{1}{6}e^{3t}$ ,

$e^{xt}$  tells us the value of  $X$  and  
the associated coefficients tell us the probabilities

(i.e.  $X=1, 2, 3$  w/ prob  $\frac{1}{2}, \frac{1}{3}, \frac{1}{6}$ .)

so MGF  $\Rightarrow$  distribution of  $X$  when  $X$  has  
finite # values,

Today

- (1) Practice with MGF
- (2) Recognizing a distribution from the variable  
part of its density
- (3) Extra Practice

# (1) Practice with MGF

For  $X \sim \text{Gamma}(r, \lambda)$

recall  $M_X(t) = \left(\frac{\lambda}{\lambda - t}\right)^r$  for  $t < \lambda$

ex Let  $X \sim \text{Exp}(\lambda)$  and  $a > 0$ .

Show that  $Y = aX$  is also exponential,  
and specify the new parameter.

Note  $M_X(s) = \left(\frac{\lambda}{\lambda - s}\right)^1$  for  $s < \lambda$

$$Y = aX$$

$$M_{aX}(t) = M_X(at) = \left(\frac{\lambda}{\lambda - at}\right)^1 = \left(\frac{\frac{\lambda}{a}}{\frac{\lambda}{a} - t}\right)^1 \quad \text{for } at < \lambda \quad \text{for } t < \frac{\lambda}{a}$$

$$\Rightarrow Y \sim \text{Exp}\left(\frac{\lambda}{a}\right)$$

or

$\begin{matrix} r & \lambda \\ \parallel & \parallel \\ \end{matrix}$

(3 pts) Let  $X_i$  follow the Gamma  $(1/100, 2/100)$  distribution for  $i = 1, 2, \dots, 100$ , independently of each other. We are interested in finding the distribution of the sample average,  $Y = \frac{1}{100} \sum_{i=1}^{100} X_i$ . Using properties of MGFs, identify the distribution of  $Y$ .

Recall that for  $X \sim \text{Gamma}(r, \lambda)$ ,  $M_X(t) = (\frac{\lambda}{\lambda - t})^r$ ,  $t < \lambda$ .

let  $S = \sum_{i=1}^{100} X_i$

$$M_S(t) = M_{X_1}(t) \cdots M_{X_{100}}(t) = \left( \left( \frac{.02}{.02 - t} \right)^{.01} \right)^{100} = \left( \frac{.02}{.02 - t} \right)^1$$

$$M_Y(t) = M_{\frac{1}{100}S}(t) = M_S(.01 t) = \frac{.02}{.02 - .01 t} = \frac{2}{2 - t}$$

Since this MGF exists in a neighborhood of zero,

By the uniqueness of MGF,  $\boxed{Y \sim \text{Exp}(2)}$

(2)

Recognizing a distribution from the variable part of its density.

A density can be written as

$$f(t) = c h(t)$$

↖ constant      ↖ variable part.

$$T_r \sim \text{Gamma}(r, \lambda), \quad r, \lambda > 0 \quad f(t) = \begin{cases} \frac{1}{\Gamma(r)} \lambda^r t^{r-1} e^{-\lambda t} & t > 0 \\ 0 & \text{else} \end{cases}$$

$$X \sim N(\mu, \sigma^2) \quad f(x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2}$$

$$1 = \int_{-\infty}^{\infty} f(t) dt = c \int_{-\infty}^{\infty} h(t) dt \Rightarrow c = \frac{1}{\int_{-\infty}^{\infty} h(t) dt}$$

So you can figure out the density from its variable part.

ex Name the distribution with the following variable part ex Gamma ( $r = \frac{1}{2}, \lambda = 3$ )

a)  $h(t) = t^3 e^{-\frac{1}{2}t}$  Gamma ( $4, \frac{1}{2}$ )

b)  $h(t) = e^{-\frac{1}{2}t^2}$  Normal ( $0, 1$ )

c)  $h(t) = e^{-3t}$  Exp (3)

d)  $h(t) = t^{-\frac{1}{2}} e^{-t}$  Gamma ( $\frac{1}{2}, 1$ )

e)  $h(t) = 1_{(t \in (0, 2))}$  Unit ( $0, 2$ )

### ③ Extra practice

Sec 4.4 Change of variable rule.

Find density of  $Y = g(X)$ ,

steps

① range of  $Y$

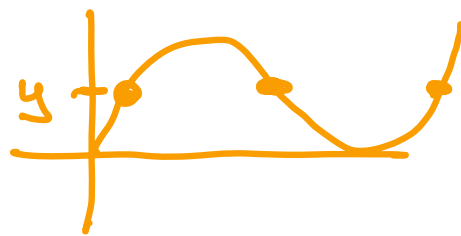
② solve for  $X$

③ Find  $\left| \frac{dx}{dy} \right|$

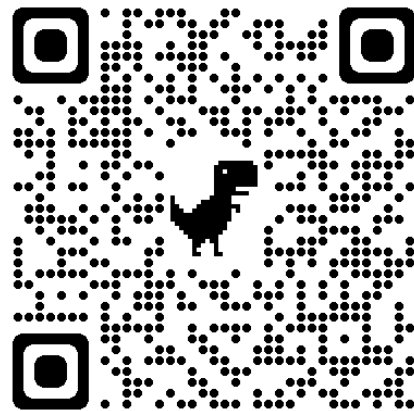
④  $f_Y(y) = \sum \left| \frac{dx}{dy} \right| f_X(x) \Big|_{x=g^{-1}(y)}$

$\# \{x: g(x) = y\}$

$\uparrow$   $n \geq 1$  above



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Let  $X$  be the standard normal RV. The distribution of  $Y = X^2$  is:

**a** Gamma( $\frac{1}{2}, \frac{1}{2}$ )

**b** Gamma( $\frac{3}{2}, \frac{1}{2}$ )

**c** Normal(0,1)

**d** none of the above

$$f(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}$$

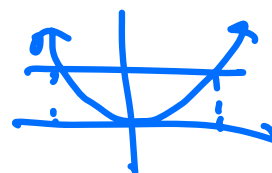
range of  $y$  is  $[0, \infty)$

$$x = \pm\sqrt{y}$$

$$\left| \frac{dx}{dy} \right| = \frac{1}{2\sqrt{y}}$$

$$f_y(y) = 2 \cdot \frac{1}{2\sqrt{y}} \cdot \frac{1}{\sqrt{2\pi}} e^{-\frac{y}{2}} = \frac{1}{\sqrt{2\pi}} e^{-\frac{y}{2}}$$

$$\# \{x : x^2 = y\}$$



every  $y$   
has  
 $2 x$  for  
 $y > 0$ .

$$y^{-1/2} e^{-y/2}$$

↑  
variable part  
of Gamma( $\frac{1}{2}, \frac{1}{2}$ )



## M6F Challenge

Can you solve this problem  
using M6F?

Hint  $X \sim N(0, \sigma^2)$  has  
density  $f(x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{x^2}{2\sigma^2}}$

Hint

M6F of Gamma( $r, \lambda$ ) is

$$\left( \frac{\lambda}{\lambda - t} \right)^r$$

So M6F of Gamma( $\frac{1}{2}, \frac{1}{2}$ )

is

$$\left( \frac{\frac{1}{2}}{\frac{1}{2} - t} \right)^{1/2}$$

$$M_{x^2}(t) = E(e^{tx^2})$$

$$= \int_{-\infty}^{\infty} e^{tx^2} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx$$

$$= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-x^2(\frac{1}{2}-t)} dx$$

need this to be  $\frac{1}{2\sigma^2}$

$$\Rightarrow \sigma^2 = \frac{1}{2(\frac{1}{2}-t)}$$

$$= \frac{1}{\sqrt{2\pi}} \cdot \sqrt{2\pi\sigma^2} = 1 \quad \text{for } t < \frac{1}{2}$$

$$= \left( \frac{\frac{1}{2}}{\frac{1}{2}-t} \right)^{\frac{1}{2}}$$

mgf of  $\text{Gamma}(\frac{1}{2}, \frac{1}{2})$

for  $t < \frac{1}{2}$