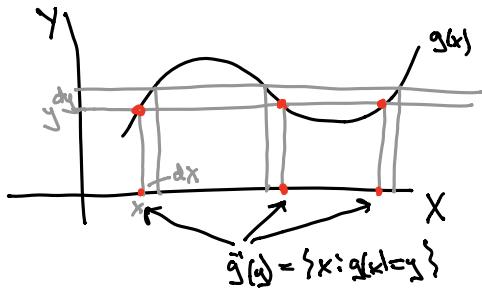


Last 2 times

(1) Sec 4.4 Change of Variable rule

many to one g:



$$\begin{aligned} f_y(y) dy &= f_x(x_1) dx_1 + f_x(x_2) dx_2 + f_x(x_3) dx_3 \\ f_y(y) &= f_x(x_1) \frac{dx_1}{dy} + f_x(x_2) \frac{dx_2}{dy} + f_x(x_3) \frac{dx_3}{dy} \\ &= \frac{f_x(x_1)}{|g'(x_1)|} + \frac{f_x(x_2)}{|g'(x_2)|} + \frac{f_x(x_3)}{|g'(x_3)|} \end{aligned}$$

P( $x \in dx_i \geq 0$ )

(2) MGF (not in book)

$$M_X(t) = E(e^{tX})$$

Thm If a MGF exists in an interval

around zero,  $M^{(k)}(t) \Big|_{t=0} = E(X^k)$

e.g.  $X \sim \text{Pois}(n)$

$$\begin{aligned} M_X(t) &= E(e^{tX}) = \sum_{k=0}^{\infty} e^{tk} \frac{n^k e^{-n}}{k!} = e^{-n} \sum_{k=0}^{\infty} \frac{(ne^t)^k}{k!} \\ &= e^{-n} e^{ne^t} = \boxed{e^{n(e^t - 1)}} \text{ for all } t \end{aligned}$$

then  $E(X) = \left. \frac{d}{dt} M_X(t) \right|_{t=0} = \left. e^{n(e^t - 1)} \cdot ne^t \right|_{t=0} = n \checkmark$

Today

(1) Key Properties of MGF

(2) Review Concept test responses from Lec 23.

(3) Sec 3.5 CDF of a mixed distribution,

# ① Key Properties of MGF

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

last time

(b) If  $X$  and  $Y$  are independent RVs,

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

Proved in MGF HW.

(c)  $M_X(t) = M_Y(t)$  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.

$\underline{\text{e.g. If }} M_X(t) = \frac{1}{2}e^t + \frac{1}{3}e^{2t} + \frac{1}{6}e^{3t},$

$e^{xt}$  tells us the value of  $X$  and

the associated coefficients tell us the probability

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

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

Property (a) is useful to find  $E(k), \text{Var}(k)$ ,

Properties (b) and (c) allow us to prove

for example that sum of independent Poisson is Poisson.

$$\stackrel{\text{ex}}{=} \left. \begin{array}{l} X_1 \sim \text{Pois}(M_1) \\ X_2 \sim \text{Pois}(M_2) \end{array} \right\} \text{independent.}$$

Show that  $X_1 + X_2 \sim \text{Pois}(M_1 + M_2)$

$$M_{X_1}(t) = e^{M_1(e^t - 1)} \quad \text{for all } t$$

$$M_{X_2}(t) = e^{M_2(e^t - 1)} \quad \text{for all } t$$

$$M_{X_1 + X_2}(t) = M_{X_1}(t)M_{X_2}(t) = \boxed{e^{(M_1 + M_2)(e^t - 1)}}$$

M6 F of  
Pois  $(M_1 + M_2)$  for all t.

$$\Rightarrow X_1 + X_2 \sim \text{Pois}(M_1 + M_2)$$

Ex Let  $X$  be a RV and  $a$  a constant.

$$\text{Show that } M_{aX}(t) = M_X(at)$$

$$\text{hint } M_{aX}(t) = E(e^{atX})$$

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

$$= M_X(at).$$

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

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

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

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

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

$$= \frac{\lambda}{(\frac{\lambda}{a}-t)} \quad \text{for } t < \left(\frac{\lambda}{a}\right)^{>0}$$

↗ MGF for  $\text{Exp}(\frac{\lambda}{a})$

By uniqueness of MGF,  $aX \sim \text{Exp}(\frac{\lambda}{a})$ .

② . Let  $V$  be a standard normal RV. The distribution of  $X = V^2$  is?

- a** Gamma
- b** Uniform
- c** Normal
- d** none of the above

a

This is a change of variable from  $V \sim \text{Norm}(0,1)$  to  $X = V^2$ . Plugging in you see that the variable part consists of  $e^{(-x/2)}/\sqrt{x}$ . This matches gamma.

a

Gamma can talk on positive real numbers that aren't usually integers, so this has to be the only option.

d

It is like a gamma distribution and gaussian distribution mixed.

c

If we have a normal distribution and square our RV then we will see that it stays normal

$$\begin{aligned}
 V &\sim N(0,1) \\
 f_V(v) &= \frac{1}{\sqrt{2\pi}} e^{-\frac{v^2}{2}} \\
 f_X(x) &= \frac{1}{\sqrt{2\pi}} e^{-\frac{v^2}{2}} \Big|_{V=\sqrt{x}} + \frac{1}{\sqrt{2\pi}} e^{-\frac{v^2}{2}} \Big|_{V=-\sqrt{x}} \\
 &\propto \frac{-x}{\sqrt{x}} = x^{-\frac{1}{2}} e^{-\frac{1}{2}x} \\
 &\text{Gamma } (\nu = \frac{1}{2}, \lambda = \frac{1}{2}x)
 \end{aligned}$$

$$f_X(x) = \frac{1}{\Gamma(\frac{1}{2})} \left(\frac{1}{2}\right)^{\frac{1}{2}} x^{-\frac{1}{2}} e^{-\frac{1}{2}x}$$

constant part,

variable part

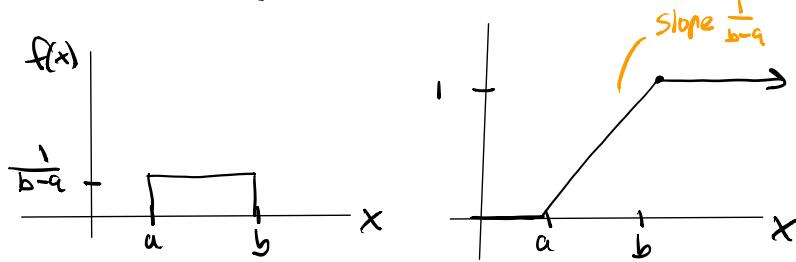
(2) Sec 4.5 Cumulative Distribution Function (CDF)

def'n  $X$  RV

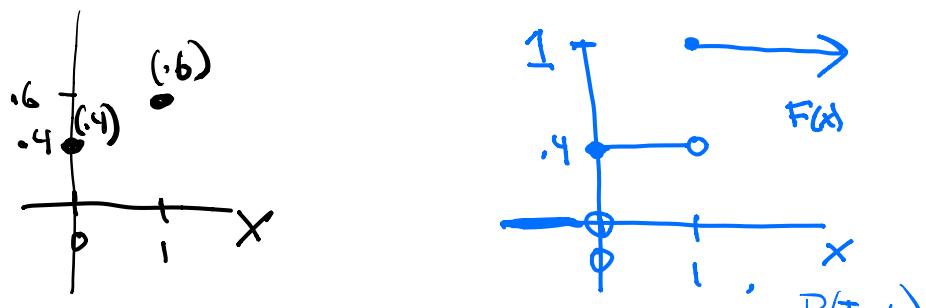
$$F_X(x) = P(X \leq x)$$

use Describes a distribution  
(equivalent to a density or probability mass function)

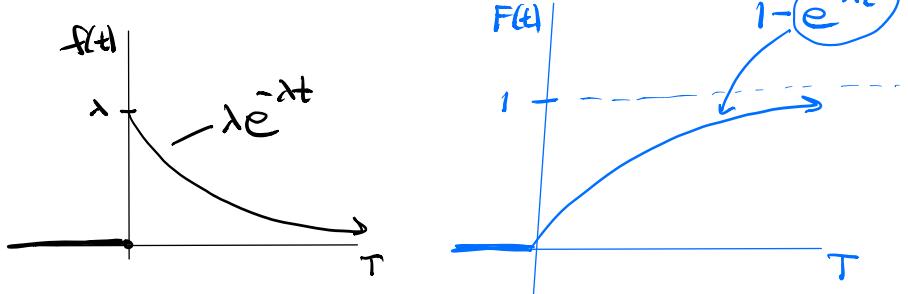
ex  $X \sim \text{Unif}(a, b)$



ex  $X \sim \text{Bernoulli}(p=0.6)$



ex  $T \sim \text{Exp}(\lambda)$



$\Leftrightarrow$  (Mixed distribution)

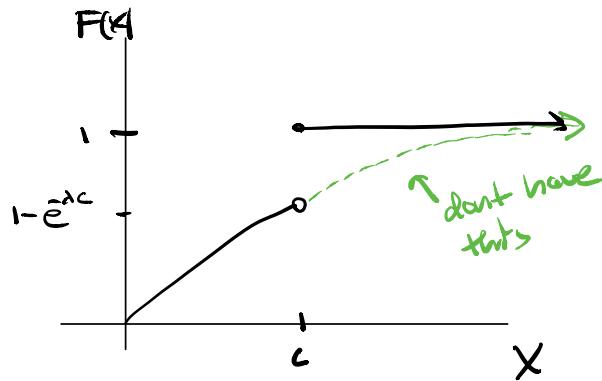
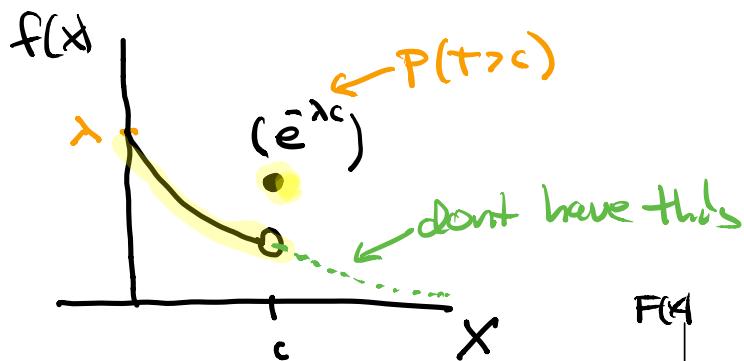
$$T \sim \text{Exp}(\lambda)$$

$$C > 0$$

$$X = \begin{cases} T & \text{if } x < C \\ C & \text{if } x = C \end{cases}$$

$$X = \min(T, C)$$

"T killed by C"



# Stat 134

## Friday March 22 2019

1. Suppose you are trying to discretely leave a party. Your time to leave is uniform from 0 to 2 minutes. However, if your walk to the exit takes more than 1 minute, you run into a friend at the door and must spend the full 2 minutes to leave. Let  $T$  represent the time it takes you to leave. True or false, the graph of the cdf of  $T$  is:

