

Last two times:

Sec 4.5 ① CDF,  $F(t) = P(T \leq t)$

useful to find distribution of min or max.

$$\stackrel{\text{ex}}{=} X_i \text{ indep } \text{Exp}(\lambda_i)$$

$$\text{let } M = \min(X_1, \dots, X_n).$$

What is distribution of  $M$ ?

$$P(M > t) = P(X_1 > t, X_2 > t, \dots, X_n > t), \text{ for } t > 0$$

$$= P(X_1 > t) \cdot P(X_2 > t) \cdots P(X_n > t)$$

$$= e^{-\lambda_1 t} \cdot e^{-\lambda_2 t} \cdots e^{-\lambda_n t}$$

$$= e^{-(\lambda_1 + \dots + \lambda_n)t}$$

$$\Rightarrow M \sim \text{expon}(\lambda_1 + \dots + \lambda_n)$$

class notes ② MGF

$X$  RV,  $t \in \mathbb{R}$ .

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

$\Leftrightarrow$  if  $X$  takes values  $X=1, 2, 3$  with prob  $\frac{1}{2}, \frac{1}{3}, \frac{1}{6}$

respectively,  $M_X(t) = \sum_{x=1}^3 e^{tx} P(x) = e^t \cdot \frac{1}{2} + e^{2t} \cdot \frac{1}{3} + e^{3t} \cdot \frac{1}{6}$

$$\stackrel{\text{ex}}{=} X \sim \text{Gamma}(r, \lambda)$$

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

for all  $t$ .

Todays

- ① Two RV with same MGF have same cdf.
- ② Review Concept test responses from lec 24.
- ③ Finding  $E(X)$  using cdf.

# ① Key Properties of MGF

(a) If an MGF exists in an interval

$$\text{containing } \mathbb{E}[X], M^{(k)}(t) \Big|_{t=0} = E(X^k)$$

(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.

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

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

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

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.

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

$$P(X_1 = k) = \frac{e^{-M_1} M_1^k}{k!}$$

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

Step 1 Find  $M_{X_1}(t) = E(e^{tX_1})$

$$E(e^{tX_1}) = \sum_{k=0}^{\infty} e^{kt} \frac{e^{-M_1} M_1^k}{k!} = e^{-M_1} \sum_{k=0}^{\infty} \frac{e^{kt} M_1^k}{k!}$$

$$= e^{-M_1} \sum_{k=0}^{\infty} \frac{(e^{tM_1})^k}{k!} = e^{-M_1} e^{tM_1}$$

$$\Rightarrow = e^{tM_1 - M_1} \quad \text{for all } t$$

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

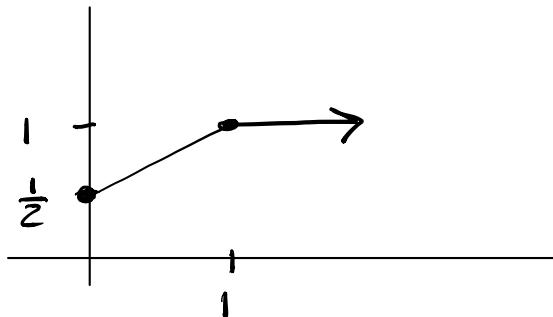
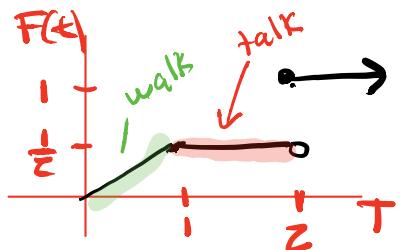
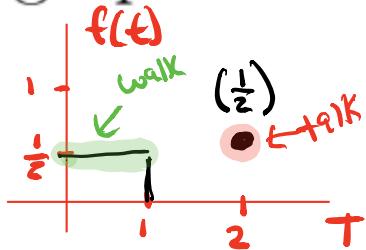
$$M_{X_2}(t) = e^{M_2(e^t - 1)} \quad \text{for all } t$$

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

↑  
M6 F of  
Pois  $(M_1+M_2)$

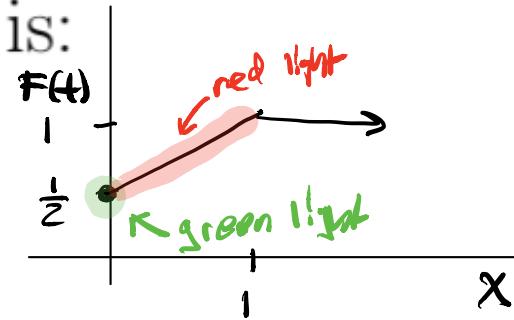
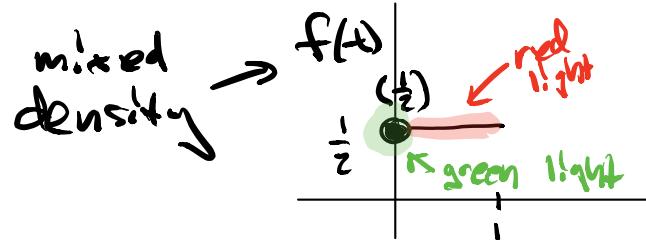
② Review concept test responses from lec 24,

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:

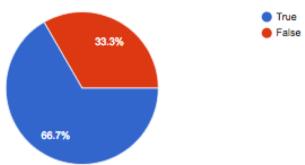


10/17/2018 9:50:36	FALSE	FALSE	should be steeper
10/17/2018 9:50:36	FALSE	FALSE	The graph shouldn't start at 1/2 that would imply that half the probability is assigned to leaving at 0. It should start at 0 and increase uniformly to t=1. Then there should be a constant line from t=1 to t<2 then a jump to 1 at t=2
10/17/2018 9:47:01	FALSE	FALSE	Increase from 0 to 1/2 between 1 and 0 and then flat from 1-2 with jump up to 1 at 2
10/17/2018 9:47:20	FALSE	FALSE	
10/17/2018 9:47:32	FALSE	FALSE	Should be constant with a jump up to one after half the time
10/17/2018 9:51:48	TRUE	FALSE	the values should start at 0 for uniform
10/17/2018 9:47:40	FALSE	FALSE	
10/17/2018 9:47:42	FALSE	FALSE	If it's from uniform from 0-2 minutes it should start at 0 and go to a half, and then jump up to one?
10/17/2018 9:47:42	FALSE	FALSE	The probability it takes you 0 minutes to leave is not 1/2 because it should reflect uniform from 0 to 2 minutes—the probability should accumulate over time starting at 0
10/17/2018 9:48:14	TRUE	FALSE	the CDF should be a flat line at 1 at 1 minute in, as at that point the time taken is guaranteed to be 2 min total. up until that point. Since the probability graph is uniform, the line before and up to 1 min in should be a flat line with a constant slope. However, the graph should start at the origin, as there is a 0% chance you can make it to the door in less than or equal to 0 sec.
10/17/2018 9:48:15	FALSE	FALSE	should start from 0
10/17/2018 9:48:21	FALSE	FALSE	should be a jump once it hits 2 minutes because if you stay any point past 1 minute, you have to take 2 minutes
10/17/2018 9:48:25	FALSE	FALSE	You shouldn't start at 1/2 and it should not be 1 until 2 min
10/17/2018 9:48:32	FALSE	FALSE	Must spend 2mins to leave so in (0, 2) p=0
10/17/2018 9:48:34	FALSE	FALSE	
10/17/2018 9:48:35	FALSE	FALSE	You aren't necessarily gone after two minutes. Also you don't have a .5 probability of already being gone by Time = 0
10/17/2018 9:50:36	FALSE	FALSE	0→1/2—1/2→1
10/17/2018 9:50:36	FALSE	FALSE	The sloped part of the function should be shifted down by 1/2, and there should be a jump discontinuity at x=1.
10/17/2018 9:54:04	FALSE	FALSE	
10/17/2018 9:54:30	FALSE	FALSE	
10/17/2018 9:54:31	FALSE	FALSE	The cdf cannot start at 1/2 because that means there is a 0.5 probability that you're already gone by the time you decide to leave
10/17/2018 9:54:31	FALSE	FALSE	Start the discontinuity from 2

Suppose stop lights at an intersection alternately show green for one minute, and red for one minute (no yellow). Suppose a car arrives at the lights at a time distributed uniformly from 0 to 2 minutes. Let  $X$  be the delay of the car at the lights (assuming there is only one car on the road). True or false, the graph of the cdf of  $X$  is:



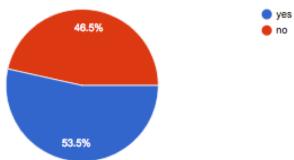
45 responses



## feedback

Did student feedback from part 1 help you?

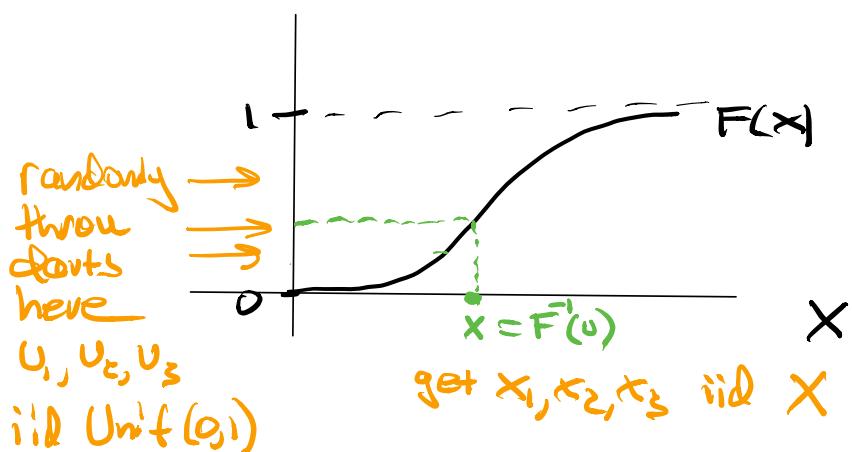
43 responses



③ Sec 4.5 Using CDF to find  $E(X)$ . sec 4.5.q

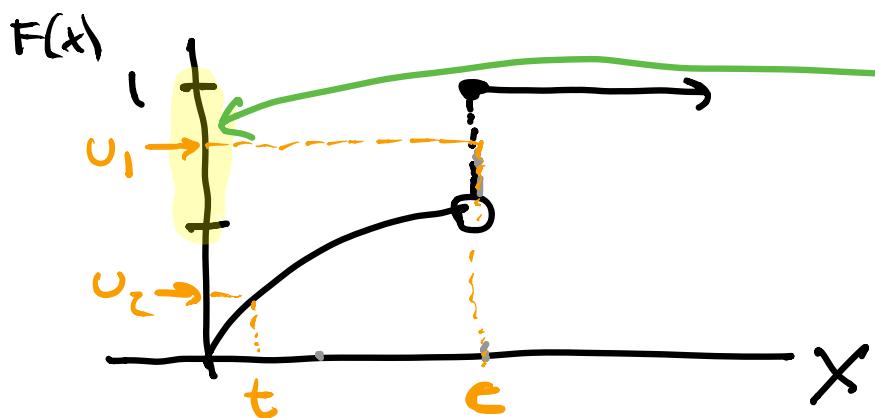
Inverse distribution function,  $F^{-1}(v)$

Let  $X$  have CDF  $F(x)$ .



\ doesn't have to be continuous

$$\Leftrightarrow X = \min(T, c), T \sim \text{exp}(1)$$



This yellow region  
gets assigned the  
single value  $c$ .

Thm (1522) — Proof at end of lecture.

Let  $X$  have CDF  $F$ .

Then the RV  $\bar{F}^{-1}(U) = X$

How is this useful to us finding  $E(X)$ ?

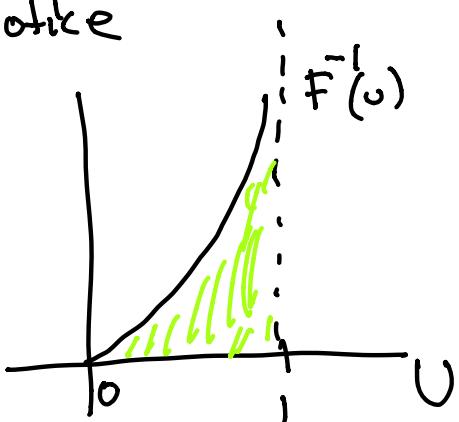
It is sometimes easier to calculate

$E(X)$  using the cdf (avoid doing  
integration by parts):

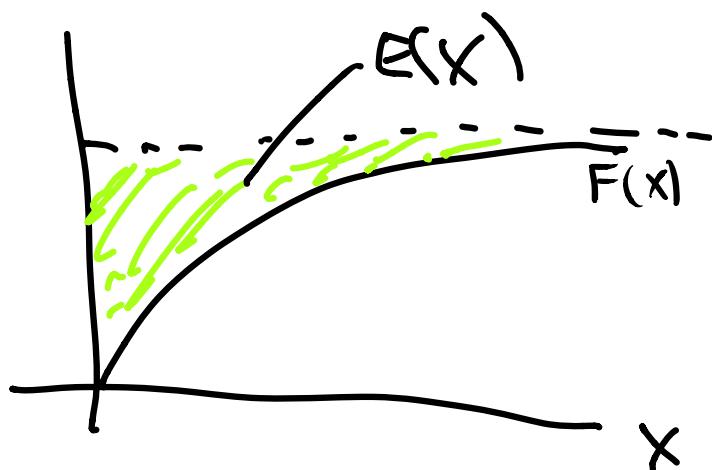
1 since  $U \sim \text{Unif}(0,1)$

$$E(X) = E(\bar{F}^{-1}(U)) = \int_0^1 \bar{F}^{-1}(U) f_U(u) du$$

Notice



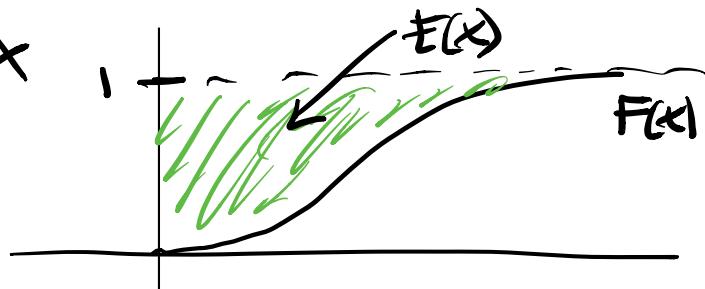
Now reflect  
the above graph  
about the  
diagonal  $y=x$



We can find the shaded region by integrating  
 $1 - F(x)$  with respect to  $x$ :

Then Let  $X$  be a pos. random variable,  
with CDF  $F$ . (continuous, discrete, mixed),

$$E(X) = \int_0^\infty (1 - F(x)) dx$$



$\text{exp}$

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

$$F_T(t) = 1 - e^{-\lambda t}$$

Calculate  $E(T)$ .

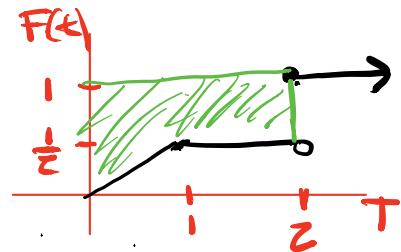
$$\begin{aligned} E(T) &= \int_0^\infty (1 - F(t)) dt = \int_0^\infty e^{-\lambda t} dt \\ &= \frac{1}{\lambda} e^{-\lambda t} \Big|_0^\infty = \boxed{\frac{1}{\lambda}} \end{aligned}$$

wow that  
was easy!

ex

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:

$$E(X) = \frac{5}{4}$$



# Concept test:

Stat 134

Monday October 21 2018

1. Let  $X$  have density  $f(x) = xe^{-x}$  for  $x > 0$ .  
The MGF is?

- a  $M_X(t) = \frac{1}{1-t}$  for  $t < 1$
- b  $M_X(t) = \frac{1}{(1-t)^2}$  for  $t < 1$
- c  $M_X(t) = \frac{1}{(1+t)^2}$  for  $t > -1$
- d none of the above

Variable part of Gamma  $x^{r-1} e^{-\lambda x}$

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

know MGF of  $\text{Gamma}(r, \lambda)$  is  $\left(\frac{\lambda}{\lambda-t}\right)^r, t < \lambda$

so 
$$M_X(t) = \frac{1}{(1-t)^r} \text{ for } t < 1$$

2. The  $MGF$  of  $X$  is  $M_X(t) = \frac{1}{\sqrt{1-t}}$  for  $t < 1$ .

The distribution of  $X$  is:

**a** Gamma( $r = 1/2, \lambda = 1$ ) and possibly another distribution.

**b** Gamma( $r = 2, \lambda = 1$ )

**c** Gamma( $r = -1/2, \lambda = 1$ )

**d** none of the above

$$X \sim \text{Gamma}(r=1/2, \lambda=1)$$

Uniquely determined from

$$M_X(t) = \frac{1}{\sqrt{1-t}} \quad \text{or} \quad t < 1.$$

## Appendix

→ See p 322 in book

Claim for any CDF  $F$

$X = F^{-1}(U)$  is a RV with cdf  $F$ .

Proof / let  $X = F^{-1}(U)$   $\leftarrow U \sim \text{Unif}(0,1)$

$$F_X(x) = P(X \leq x) \quad \text{we will show} \quad F_X = F$$

$$= P(F^{-1}(U) \leq x)$$

$$= P(F(F^{-1}(U)) \leq F(x)) \quad \text{Since } F \text{ is increasing}$$

$$= P(U \leq F(x))$$

$$= F(x) \quad \text{since } P(U \leq v) = v$$

□