

# MATH 180A: INTRO TO PROBABILITY (FOR DATA SCIENCE)

[www.math.ucsd.edu/~tkemp/180A](http://www.math.ucsd.edu/~tkemp/180A)

Today: § 5.1-5.2

Next: § 6.1-6.2

Lab 5 due Wednesday (Nov 13) by 11:59 pm

HW 5 due **TONIGHT** by 11:59 pm

↳ Error in problem statement of Exercise 4.40:  
the exact probability is  $\approx 0.00327556$   
 $\neq 0.000949681$

# Functions to Describe Probability Distributions

5.1

Random variable  $X$ .

\* CDF  $F_X(t) = P(X \leq t)$

- Works every time
- But can be hard to compute.
- No clear relation to  $E(X)$

\* PMF  $P_X(k) = P(X = k)$

- Only when  $X$  is discrete,
- $E(X) = \sum_k k P_X(k)$

\* PDF  $f_X(t) = \frac{d}{dt} F_X(t)$

- Only when  $X$  is continuous.
- $E(X) = \int_{-\infty}^{\infty} x f_X(x) dx$

New entry: MGF Moment Generating Function.

$$M_X(t) = E(e^{tx}) \begin{cases} = \sum_k e^{tk} P_X(k) & \text{if } X \text{ discrete} \\ = \int_{-\infty}^{\infty} e^{tx} f_X(x) dx & \text{if } X \text{ contin.} \end{cases}$$

E.g.  $X \sim \text{Ber}(p)$   $P(X=1) = p$   $\therefore M_X(t) = \mathbb{E}(e^{tX}) = e^{t \cdot 0} (1-p) + e^{t \cdot 1} (p)$   
 $P(X=0) = 1-p$   $p(e^t - 1) + 1 \leftarrow = e^t p + (1-p)$

E.g.  $N \sim \text{Poisson}(\lambda)$   $P(N=k) = e^{-\lambda} \frac{\lambda^k}{k!}$   
 $M_N(t) = \mathbb{E}(e^{tN}) = \sum_{k=0}^{\infty} e^{tk} e^{-\lambda} \frac{\lambda^k}{k!} = e^{-\lambda} \sum_{k=0}^{\infty} \frac{(e^t \lambda)^k}{k!} = e^{-\lambda} e^{e^t \lambda}$   
 $(e^t)^k \lambda^k = (e^t \lambda)^k$   $= e^{e^t \lambda - \lambda}$   
 $= e^{\lambda(e^t - 1)}$

E.g.  $Z \sim \mathcal{N}(0,1)$

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

$$\left[ -\frac{x^2}{2} + tx = -\frac{1}{2}(x^2 - 2tx) = -\frac{1}{2}(x^2 - 2tx + t^2 - t^2) = -\frac{1}{2} \underbrace{(x-t)^2}_{(x-t)^2} + \frac{t^2}{2} \right]$$

$$\rightarrow M_Z(t) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \underbrace{e^{-\frac{1}{2}(x-t)^2}}_{e^{-\frac{1}{2}(x-t)^2}} \cdot \underbrace{e^{\frac{t^2}{2}}}_{e^{\frac{t^2}{2}}} dx = e^{\frac{t^2}{2}} \cdot \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-(x-t)^2/2} dx = e^{\frac{t^2}{2}}$$

$$\underbrace{\frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-u^2/2} du}_{= 1} = 1$$

E.g.  $T \sim \text{Exp}(\lambda)$

$$M_T(\xi) = \mathbb{E}(e^{\xi T}) = \int_0^{\infty} e^{\xi s} \cdot \lambda e^{-\lambda s} ds$$

$$f_T(s) = \begin{cases} \lambda e^{-\lambda s} & s > 0 \\ 0 & s \leq 0 \end{cases}$$

$$= \lambda \int_0^{\infty} e^{\xi s - \lambda s} ds$$

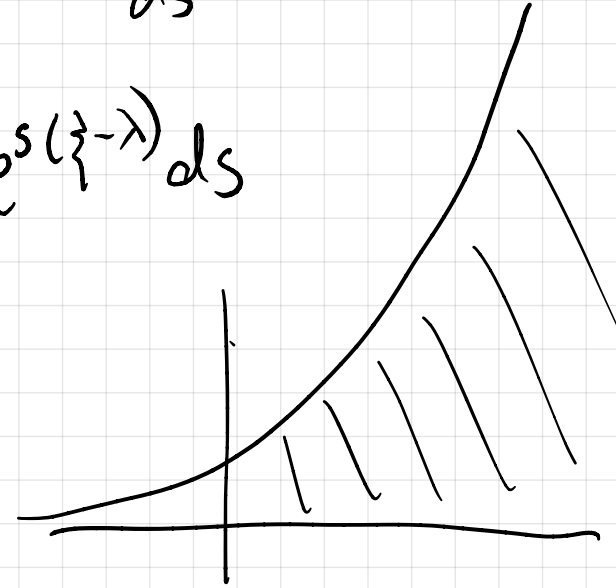
$$= \lambda \int_0^{\infty} e^{s(\xi - \lambda)} ds$$

$$M_T(\xi) = \infty \text{ if } \xi \geq \lambda$$

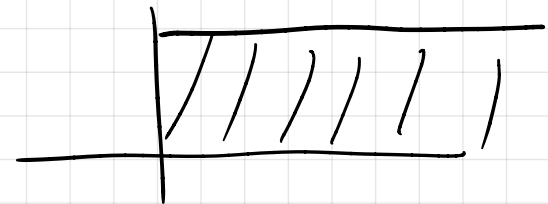
$$= \lambda \cdot \frac{1}{b}$$

$$= \frac{\lambda}{\lambda - \xi} \text{ if } \xi < \lambda$$

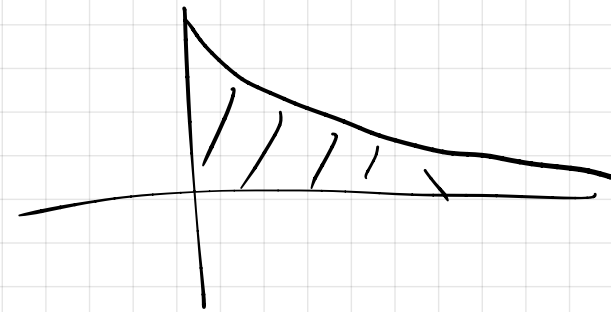
If  $\xi - \lambda > 0$



If  $\xi - \lambda = 0$



If  $\xi - \lambda < 0$



If  $b = \xi - \lambda < 0$

$$\begin{aligned} & \int_0^{\infty} e^{bs} ds \\ &= \frac{1}{b} e^{bs} \Big|_0^{\infty} \\ &= \frac{1}{b} (0 - 1) \end{aligned}$$

A MGF may take some infinite values.  
 There is always at least one finite value:

$$M_X(0) = \mathbb{E}(e^{0 \cdot X}) = \mathbb{E}(1) = 1.$$

But it can happen that there are no others!

Eg. Cauchy density  $f(x) = \frac{1}{\pi(1+x^2)} \sim X$



$$\int_{-\infty}^{\infty} f(x) dx = \frac{1}{\pi} \int_{-\infty}^{\infty} \frac{1}{1+x^2} dx$$

$$= \frac{1}{\pi} \tan^{-1}(x) \Big|_{-\infty}^{\infty} = \frac{1}{\pi} \left( \frac{\pi}{2} - \left( -\frac{\pi}{2} \right) \right) = 1.$$

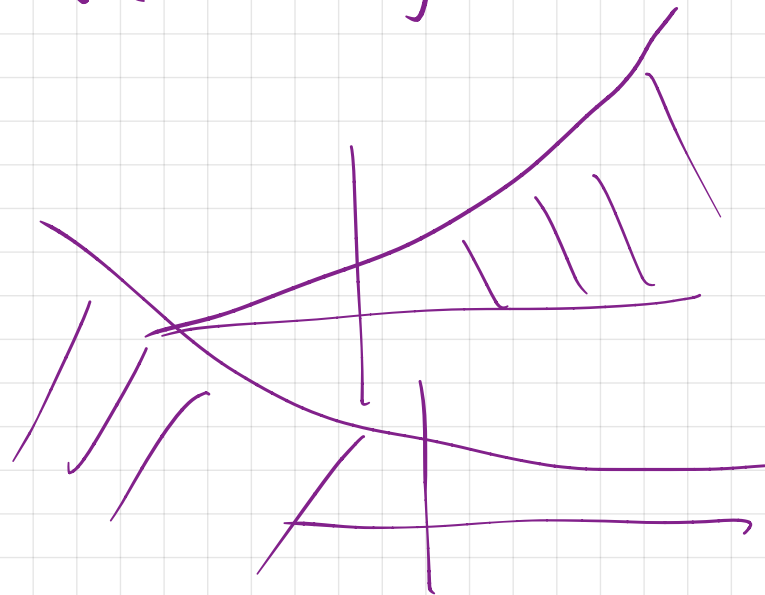
$$\mathbb{E}(e^{tX}) = \frac{1}{\pi} \int e^{tx} \cdot \frac{1}{1+x^2} dx$$

If  $t > 0$ ,

If  $t < 0$ ,

$$\frac{e^{tx}}{1+x^2}$$

$$\frac{e^{tx}}{1+x^2}$$



# Why MGF?

Given a random variable  $X$ , its moments (should they exist) are the numbers  $\mathbb{E}(X^k)$ ,  $k=0,1,2,\dots$

These can be computed from  $M_X(t)$  as follows.

$$\frac{d}{dt} \mathbb{E}(e^{tX}) = \mathbb{E}\left(\frac{d}{dt} e^{tX}\right) = \mathbb{E}(Xe^{tX})$$

$$\therefore \left(\frac{d}{dt} \mathbb{E}(e^{tX})\right) \Big|_{t=0} = \mathbb{E}(Xe^{0 \cdot X}) = \mathbb{E}(X) \leftarrow \text{mean}$$

⋮

$$\frac{d^k}{dt^k} M_X(0) = \mathbb{E}(X^k).$$

Theorem: Suppose  $M_X(t) < \infty$  for all  $t$  in some neighborhood of 0  
 $(-\xi, \xi)$   $\xi > 0$ .

Then  $M_X$  is analytic on this neighborhood: its Taylor series based @ 0 converges to  $M_X(t)$  on this interval, and

$$M_X(t) = \sum_{k=0}^{\infty} \frac{\mathbb{E}(X^k)}{k!} t^k$$

Eg. Find the moments of the  $\text{Exp}(\lambda)$  distribution.

$$M_T(t) = \frac{\lambda}{\lambda - t} \quad (t < \lambda)$$

$$= \frac{1}{1 - t/\lambda} = \sum_{k=0}^{\infty} \left(\frac{t}{\lambda}\right)^k = \sum_{k=0}^{\infty} \frac{1}{\lambda^k} t^k \quad \left|\frac{t}{\lambda}\right| < 1$$

$$\mathbb{E}(T^k) = \frac{k!}{\lambda^k}$$

$$\therefore \frac{\mathbb{E}(T^k)}{k!} = \frac{1}{\lambda^k}$$

$$\mathbb{E}(T) = \frac{1}{\lambda} \quad \mathbb{E}(T^2) = \frac{2}{\lambda^2} \quad \text{Var } T = \mathbb{E}(T^2) - \mathbb{E}(T)^2 = \frac{2}{\lambda^2} - \left(\frac{1}{\lambda}\right)^2 = \frac{1}{\lambda^2}$$

Eg. Find the moments of the  $\mathcal{N}(0,1)$  distribution.

$$\begin{aligned}
 M_Z(t) &= e^{t^2/2} = \sum_{k=0}^{\infty} \frac{(t^2/2)^k}{k!} = \sum_{k=0}^{\infty} \frac{1}{2^k k!} t^{2k} \\
 &= \sum_{n=0}^{\infty} \frac{\mathbb{E}(Z^n)}{n!} t^n = \sum_{\substack{n=2k \\ k=0}}^{\infty} \frac{\mathbb{E}(X^{2k})}{(2k)!} t^{2k} \\
 &\quad \uparrow \\
 &\quad \mathbb{E}(Z^n) = 0 \text{ if } n \text{ is odd}
 \end{aligned}$$

$$\therefore \frac{\mathbb{E}(X^{2k})}{(2k)!} = \frac{1}{2^k k!} \quad \therefore \mathbb{E}(X^{2k}) = \frac{(2k)!}{2^k k!}$$

# pairings of  
2k things.

$$\begin{aligned}
 & \frac{\cancel{2k} (2k-1) \cancel{(2k-2)} (2k-3) \cancel{(2k-4)} \dots (3) \cancel{(2)} (1)}{\cancel{2k} \cancel{(2k-1)} \cancel{(2k-2)} \dots \cancel{(2)} (1)} \\
 & \rightarrow = (2k-1)(2k-3)(2k-5) \dots (3)(1) \\
 & = (2k-1)!!
 \end{aligned}$$