

## Continuous Probability

CS70 Summer 2016 - Lecture 6A

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UC Berkeley

## Continuous Probability

### Logistics

Tutoring Sections - M/W 5-8PM in 540 Cory.

- Conceptual discussions of material
- No homework discussion (take that to OH/HW party, please)

Midterm is this Friday - 11:30-1:30, same rooms as last time.

- Covers material from MT1 to this Wednesday...
- ...but we will expect you to know everything we've covered from the start of class.
- One **double**-sided sheet of notes allowed (our advice: reuse sheet from MT1 and add MT2 topics to the other side).
- Students with time conflicts and DSP students should have been contacted by us - if you are one and you haven't heard from us, get in touch ASAP.

1

### Motivation I

Sometimes you can't model things discretely. Random real numbers. Points on a map. Time.

Probability space is **continuous**.

What is probability? Function mapping events to  $[0, 1]$ .

What is an event in continuous probability?

3

### Today

- What is continuous probability?
- Expectation and variance in the continuous setting.
- Some common distributions.

2

### Motivation II

Class starts at 14:10. You take your seat at some "uniform" random time between 14:00 and 14:10.

What's an event here? Probability of coming in at exactly 14:03:47.32?

Sample space: all times between 14:00 and 14:10.

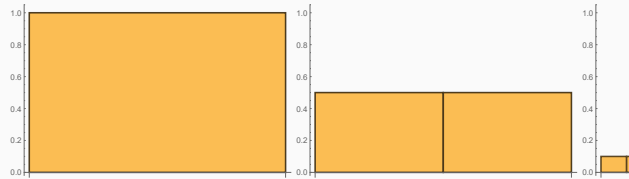
Size of sample space? How many numbers are there between 0 and 10? **infinite**

Chance of getting one event in an infinite sized uniform sample space? **0**

**Not so simple to define events in continuous probability!**

4

### Motivation III



Look at *intervals* instead of specific times.

Probability that you come in between 14:00 and 14:10? **1**.

Probability that you come in between 14:00 and 14:05? **1/2**.

Probability that you come between 14:03 and 14:04? **1/10**.

Probability that you come in some time interval of  $10/k$  minutes?  
 **$1/k$** .

5

### CDF

**Cumulative distribution function (CDF):**  $F_X(t) = \Pr[X \leq t]$ .

Or, in terms of PDF...

$$F_X(t) = \int_{-\infty}^t f_X(z) dz$$

$$\begin{aligned} \Pr[X \in (a, b]] &= \Pr[X \leq b] - \Pr[X \leq a] \\ &= F_X(b) - F_X(a) \end{aligned}$$

$$F_X(t) \in [0, 1]$$

$$\lim_{t \rightarrow -\infty} F_X(t) = 0$$

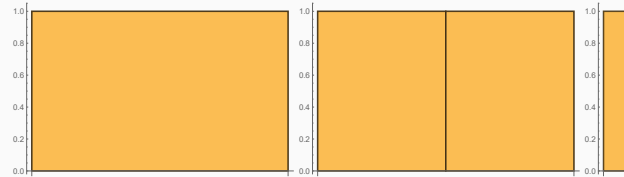
$$\lim_{t \rightarrow \infty} F_X(t) = 1$$

8

### PDF (no, not the file format)

What happens when you take  $k \rightarrow \infty$ ? Probability goes to 0.

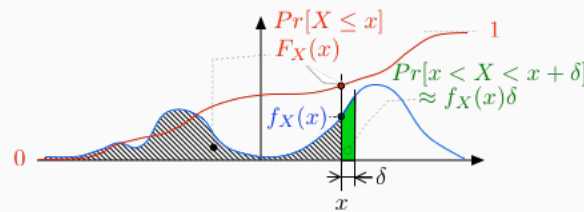
What do we do so that this doesn't disappear? **If we split our sample space into  $k$  pieces - multiply each one by  $k$ .**



The resulting curve as  $k \rightarrow \infty$  is the **probability density function (PDF)**.

6

### In Pictures



9

### Formally speaking...

PDF  $f_X(t)$  of a random variable  $X$  is defined so that the probability of  $X$  taking on a value in  $[t, t + \delta]$  is  $\delta f(t)$  for infinitesimally small  $\delta$ .

$$f_X(t) = \lim_{\delta \rightarrow 0} \frac{\Pr[X \in [t, t + \delta]]}{\delta}$$

Another way of looking at it:

$$\Pr[X \in [a, b]] = \int_a^b f_X(t) dt$$

$f$  is nonnegative (negative probability doesn't make much sense).

Total probability is 1:  $\int_{-\infty}^{\infty} f_X(t) dt = 1$

7

### Expectation

Discrete case:  $E[X] = \sum_{t=-\infty}^{\infty} (\Pr[X = t]t)$

Continuous case? Sum  $\rightarrow$  integral.

$$E[X] = \int_{-\infty}^{\infty} t f_X(t) dt$$

Expectation of a function?

$$E[g(X)] = \int_{-\infty}^{\infty} g(t) f_X(t) dt$$

Linearity of expectation:

$$E[aX + bY] = aE[X] + bE[Y]$$

Proof: similar to discrete case.

If  $X, Y, Z$  are mutually independent, then  $E[XYZ] = E[X]E[Y]E[Z]$ .

Proof: also similar to discrete case.

Exercise: try proving these yourself.

10

## Variance

Variance is defined exactly like it is for the discrete case.

$$\begin{aligned}\text{Var}(X) &= E[(X - E[X])^2] \\ &= E[X^2] - E[X]^2\end{aligned}$$

The standard properties for variance hold in the continuous case as well.

$$\text{Var}(aX) = a^2 \text{Var}(X)$$

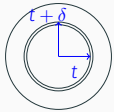
For independent r.v.  $X, Y$ :

$$\text{Var}(X + Y) = \text{Var}(X) + \text{Var}(Y)$$

11

## Target shooting III

Another way of attacking the same problem: what's the probability of hitting some ring with inner radius  $t$  and outer radius  $t + \delta$  for small  $\delta$ ?



Area of circle:  $\pi$

Area of ring:

$$\pi((t + \delta)^2 - t^2) = \pi(t^2 + 2t\delta + \delta^2 - t^2) = \pi(2t\delta + \delta^2) \approx \pi 2t\delta$$

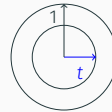
Probability of hitting the ring:  $2t\delta$ .

PDF for  $t \leq 1$ :  $2t$

14

## Target shooting

Suppose an archer always hits a circular target with 1 meter radius, and the exact point that he hits is distributed uniformly across the target. What is the distribution the distance between his arrow and the center (call this r.v.  $X$ )?



Probability that arrow is closer than  $t$  to the center?

$$\begin{aligned}\Pr[X \leq t] &= \frac{\text{area of small circle}}{\text{area of dartboard}} \\ &= \frac{\pi t^2}{\pi} = t^2.\end{aligned}$$

12

## Shifting & Scaling

Let  $f_X(x)$  be the pdf of  $X$  and  $Y = a + bX$  where  $b > 0$ . Then

$$\begin{aligned}\Pr[Y \in (y, y + \delta)] &= \Pr[a + bX \in (y, y + \delta)] \\ &= \Pr[X \in (\frac{y-a}{b}, \frac{y+\delta-a}{b})] \\ &= \Pr[X \in (\frac{y-a}{b}, \frac{y-a}{b} + \frac{\delta}{b})] \\ &= f_X(\frac{y-a}{b}) \frac{\delta}{b}.\end{aligned}$$

Left-hand side is  $f_Y(y)\delta$ . Hence,

$$f_Y(y) = \frac{1}{b} f_X(\frac{y-a}{b}).$$

15

## Target shooting II

CDF:

$$F_Y(t) = \Pr[Y \leq t] = \begin{cases} 0 & \text{for } t < 0 \\ t^2 & \text{for } 0 \leq t \leq 1 \\ 1 & \text{for } t > 1 \end{cases}$$

PDF?

$$f_Y(t) = F_Y(t)' = \begin{cases} 2t & \text{for } 0 \leq t \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

13

## Continuous Distributions

## Uniform Distribution: CDF and PDF

PDF is constant over some interval  $[a, b]$ , zero outside the interval.  
What's the value of the constant in the interval?

$$\int_{-\infty}^{\infty} k dt = \int_a^b k dt = b - a = 1$$

so PDF is  $1/(b - a)$  in  $[a, b]$  and 0 otherwise.

CDF?

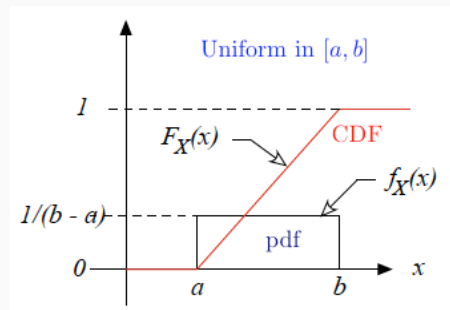
$$\int_{-\infty}^t 1/(b - a) dz$$

0 for  $t < a$ ,  $(t - a)/(b - a)$  for  $a < t < b$ , and 1 for  $t > b$ .

16

## Uniform Distribution: CDF and PDF, Graphically

$$f_X(t) = \begin{cases} 1/(b - a) & a < t < b \\ 0 & \text{otherwise} \end{cases} \quad F_X(t) = \begin{cases} 0 & t < a \\ (t - a)/(b - a) & a < t < b \\ 1 & b < t \end{cases}$$



17

## Uniform Distribution: Expectation and Variance

Expectation?

$$E[X] = \int_a^b \frac{t}{b-a} dt = \frac{1}{2} \frac{b^2 - a^2}{b-a} = \frac{b+a}{2}$$

Variance?

$$\begin{aligned} \text{Var}[X] &= E[X^2] - E[X]^2 \\ &= \int_a^b \frac{t^2}{b-a} dt - \left(\frac{b+a}{2}\right)^2 \\ &= \frac{t^3}{3(b-a)} \Big|_a^b - \left(\frac{b+a}{2}\right)^2 \\ &= \frac{(a-b)^2}{12} \end{aligned}$$

18

## Exponential Distribution: Motivation

Continuous-time analogue of the geometric distribution.

How long until a server fails? How long does it take you to run into pokemon?

Can't "continuously flip a coin". What do we do?

Look at geometric distributions representing processes with higher and higher granularity.

19

## Exponential Distribution: Motivation II

Suppose a server fails with probability  $\lambda$  every day.

Probability that server fails on the same day as time  $t$ :

$$(1 - \lambda)^{\lceil t \rceil - 1} \lambda$$

More precision! What's the probability that it fails in a 12-hour period?  $\lambda/2$  if we assume that there is no time that it's more likely to fail than another.

Generally: server fails with probability  $\lambda/n$  during any  $1/n$ -day time period.

Probability that server fails on the same  $1/n$ -day time period as  $t$ :

$$\left(1 - \frac{\lambda}{n}\right)^{\lceil tn \rceil - 1} \frac{\lambda}{n}$$

20

## Exponential Distribution: Motivation III

$$\left(1 - \frac{\lambda}{n}\right)^{\lceil tn \rceil - 1} \frac{\lambda}{n}$$

What happens when we try to take  $n \rightarrow \infty$ ?

Probability goes to zero...but we can make a PDF out of this!

Remove the width of the interval ( $1/n$ ) and take the limit as  $n \rightarrow \infty$  to get:

$$\begin{aligned} \lim_{n \rightarrow \infty} \left(1 - \frac{\lambda}{n}\right)^{\lceil tn \rceil - 1} \frac{\lambda}{n} &= \lambda \lim_{n \rightarrow \infty} \left(1 - \frac{\lambda}{n}\right)^{tn-1} \\ &= \lambda e^{-\lambda t} \end{aligned}$$

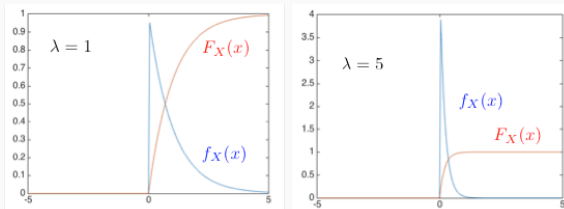
This is the PDF of the **exponential distribution**!

21

## Exponential Distribution: PDF and CDF

The exponential distribution with parameter  $\lambda > 0$  is defined by

$$f_X(t) = \begin{cases} 0, & \text{if } t < 0 \\ \lambda e^{-\lambda t}, & \text{if } t \geq 0. \end{cases} \quad F_X(t) = \begin{cases} 0, & \text{if } t < 0 \\ 1 - e^{-\lambda t}, & \text{if } t \geq 0. \end{cases}$$



Note that  $\Pr[X > t] = e^{-\lambda t}$  for  $t > 0$ .

22

## Properties of the Exponential Distribution: Scaling

Let  $X = \text{Expo}(\lambda)$  and  $Y = aX$  for some  $a > 0$ . Then

$$\begin{aligned} \Pr[Y > t] &= \Pr[aX > t] = \Pr[X > t/a] \\ &= e^{-\lambda(t/a)} = e^{-(\lambda/a)t} = \Pr[Z > t] \text{ for } Z = \text{Expo}(\lambda/a). \end{aligned}$$

Thus,  $a \times \text{Expo}(\lambda) = \text{Expo}(\lambda/a)$ . Also,  $\text{Expo}(\lambda) = \frac{1}{\lambda} \text{Expo}(1)$ .

25

## Expectation & Variance of the Exponential Distribution

$X = \text{Expo}(\lambda)$ . Then,  $f_X(x) = \lambda e^{-\lambda x} 1\{x \geq 0\}$ . Thus,

$$E[X] = \int_0^\infty x \lambda e^{-\lambda x} dx = - \int_0^\infty x d e^{-\lambda x}.$$

Integration by parts:

$$\begin{aligned} \int_0^\infty x d e^{-\lambda x} &= [x e^{-\lambda x}]_0^\infty - \int_0^\infty e^{-\lambda x} dx \\ &= 0 - 0 + \frac{1}{\lambda} \int_0^\infty d e^{-\lambda x} = - \frac{1}{\lambda}. \end{aligned}$$

So: expectation is  $E[X] = \frac{1}{\lambda}$ .

Variance:  $1/\lambda^2$

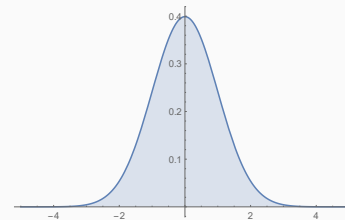
23

## Normal Distribution

Continuous counterpart to Binomial dist. (more on this later)

**Normal (or Gaussian) distribution** with parameters  $\mu, \sigma^2$ , denoted  $\mathcal{N}(\mu, \sigma^2)$ :

$$f_X(t) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(t-\mu)^2}{2\sigma^2}}$$



Sometimes called a "bell curve". Above:  $\mathcal{N}(0, 1)$ .

26

## Properties of the Exponential Distribution: Memorylessness

Similar to memorylessness for geometric distributions.

"If your server doesn't fail today, it's in the same state as it was before today."

Let  $X = \text{Expo}(\lambda)$ . Then, for  $s, t > 0$ ,

$$\begin{aligned} \Pr[X > t + s \mid X > s] &= \frac{\Pr[X > t + s]}{\Pr[X > s]} \\ &= \frac{e^{-\lambda(t+s)}}{e^{-\lambda s}} = e^{-\lambda t} \\ &= \Pr[X > t]. \end{aligned}$$

24

## Normal Distribution: Properties

$$\text{PDF: } f_X(t) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(t-\mu)^2}{2\sigma^2}}$$

CDF: involves an integral with no nice closed form (often expressed in terms of "erf", the error function). Won't discuss it here.

Expectation:  $\mu$  (notice that PDF is symmetric around  $\mu$ ).

Variance:  $\sigma^2$  (fairly straightforward integration)

Scaling/Shifting: if  $X \sim \mathcal{N}(0, 1)$  and  $Y = \mu + \sigma X$ , then  $Y \sim \mathcal{N}(\mu, \sigma^2)$ .

27

## Central Limit Theorem

Basically: if you take a lot of i.i.d random variables from any\* distribution with zero mean and the same variance and sum them up, the sum will converge to a random Gaussian with the same mean and variance.

Suppose  $X_1, X_2, \dots$  are i.i.d. random variables with expectation  $\mu$  and variance  $\sigma^2$ . Let

$$S_n := \frac{A_n - \mu}{\sigma/\sqrt{n}} = \frac{(\sum_i X_i) - n\mu}{\sigma/\sqrt{n}}$$

Then  $S_n$  tends towards  $\mathcal{N}(0, 1)$  as  $n \rightarrow \infty$ .

Or:

$$\Pr[S_n \leq a] \rightarrow \frac{1}{\sqrt{2\pi}} \int_{-\infty}^a e^{-x^2/2} dx$$

Proof: EE126

Sum of Bernoullis (binomial) tends towards normal!

28

## Summary

Continuous probability: translation of discrete probability to a continuous sample space with an infinite number of events.

Concepts of variance, expectation, etc. translate to continuous too.

Geometric distribution  $\rightarrow$  exponential distribution.

Binomial distribution  $\rightarrow$  normal distribution.

Central limit theorem: everything converges to normal if we take enough samples

29

## Today's Gig: Cauchy Distribution

## Cauchy



**Augustin-Louis Cauchy (1789-1857)**

Practically invented complex analysis. Made fundamental contributions to calculus and group theory.

"More concepts and theorems have been named for Cauchy than for any other mathematician."

Was also a baron because he tutored a duke... who ended up hating math.

30

## Definition

Actually first written about by Poisson in 1824. Cauchy became associated with it in 1853!

Suppose I have a wall on the x-axis. Stand at (0,1) and point a laser at a uniform random angle such that the laser hits the wall.

What is the distribution of the point on the wall?

$$\tan \theta = t$$

$$\theta = \tan^{-1} t$$

$$d\theta = \frac{1}{1+t^2} dt$$

$$\frac{d\theta}{\pi} = \frac{1}{1+t^2} \frac{dt}{\pi}$$

31

## Properties

PDF:

$$\frac{1}{\pi(1+t^2)}$$

Expectation?

$$\begin{aligned} \int_{-\infty}^{\infty} \frac{t}{\pi(1+t^2)} dt &= \lim_{a \rightarrow \infty} \int_{-a}^a \frac{t}{\pi(1+t^2)} dt = 0 \\ &= \lim_{a \rightarrow \infty} \int_{-a}^{2a} \frac{t}{\pi(1+t^2)} dt \neq 0 \end{aligned}$$

Expectation doesn't exist!

If you try to estimate the expectation by sampling points and averaging, you'll get crazy results.

Variance doesn't exist either.

Main takeaway: there are some really badly-behaved distributions out there.

32

Questions?