

# CS70: Lecture 30

## Continuous Probability (contd.)

1. Review: CDF, PDF
2. Review: Expectation
3. Review: Independence
4. Meeting at a Restaurant
5. Breaking a Stick
6. Maximum of Exponentials
7. Geometric and Exponential

## Review: CDF and PDF.

Key idea: For a continuous RV,  $Pr[X = x] = 0$  for all  $x \in \mathfrak{R}$ .

Examples: Uniform in  $[0, 1]$ ; throw a dart in a target.

Thus, one cannot define  $Pr[\text{outcome}]$ , then  $Pr[\text{event}]$ .

Instead, one **starts** by defining  $Pr[\text{event}]$ .

Thus, one defines  $Pr[X \in (-\infty, x]] = Pr[X \leq x] =: F_X(x), x \in \mathfrak{R}$ .

Then, one defines  $f_X(x) := \frac{d}{dx} F_X(x)$ .

Hence,  $f_X(x)\varepsilon = Pr[X \in (x, x + \varepsilon)]$ .

$F_X(\cdot)$  is the **cumulative distribution function** (CDF) of  $X$ .

$f_X(\cdot)$  is the **probability density function** (PDF) of  $X$ .

# Expectation

**Definitions:** (a) The **expectation** of a random variable  $X$  with pdf  $f(x)$  is defined as

$$E[X] = \int_{-\infty}^{\infty} xf_X(x)dx.$$

(b) The expectation of a function of a random variable is defined as

$$E[h(X)] = \int_{-\infty}^{\infty} h(x)f_X(x)dx.$$

(c) The expectation of a function of multiple random variables is defined as

$$E[h(\mathbf{X})] = \int \cdots \int h(\mathbf{x})f_{\mathbf{X}}(\mathbf{x})dx_1 \cdots dx_n.$$

**Justifications:** Think of the discrete approximations of the continuous RVs.

# Multiple Continuous Random Variables

One defines a pair  $(X, Y)$  of continuous RVs by specifying  $f_{X,Y}(x, y)$  for  $x, y \in \mathfrak{R}$  where

$$f_{X,Y}(x, y) dx dy = Pr[X \in (x, x + dx), Y \in (y, y + dy)].$$

The function  $f_{X,Y}(x, y)$  is called the **joint pdf** of  $X$  and  $Y$ .

**Example:** Choose a point  $(X, Y)$  uniformly in the set  $A \subset \mathfrak{R}^2$ . Then

$$f_{X,Y}(x, y) = \frac{1}{|A|} \mathbf{1}\{(x, y) \in A\}$$

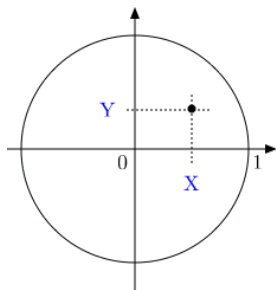
where  $|A|$  is the area of  $A$ .

**Interpretation.** Think of  $(X, Y)$  as being discrete on a grid with mesh size  $\varepsilon$  and  $Pr[X = m\varepsilon, Y = n\varepsilon] = f_{X,Y}(m\varepsilon, n\varepsilon)\varepsilon^2$ .

**Extension:**  $\mathbf{X} = (X_1, \dots, X_n)$  with  $f_{\mathbf{X}}(\mathbf{x})$ .

## Example of Continuous $(X, Y)$

Pick a point  $(X, Y)$  uniformly in the unit circle.



Thus,  $f_{X,Y}(x,y) = \frac{1}{\pi} 1\{x^2 + y^2 \leq 1\}$ .

Consequently,

$$Pr[X > 0, Y > 0] = \frac{1}{4}$$

$$Pr[X < 0, Y > 0] = \frac{1}{4}$$

$$Pr[X^2 + Y^2 \leq r^2] = r^2$$

$$Pr[X > Y] = \frac{1}{2}.$$

## Independent Continuous Random Variables

**Definition:** The continuous RVs  $X$  and  $Y$  are independent if

$$Pr[X \in A, Y \in B] = Pr[X \in A]Pr[Y \in B], \forall A, B.$$

**Theorem:** The continuous RVs  $X$  and  $Y$  are independent if and only if

$$f_{X,Y}(x,y) = f_X(x)f_Y(y).$$

**Proof:** As in the discrete case.

**Definition:** The continuous RVs  $X_1, \dots, X_n$  are mutually independent if

$$Pr[X_1 \in A_1, \dots, X_n \in A_n] = Pr[X_1 \in A_1] \cdots Pr[X_n \in A_n], \forall A_1, \dots, A_n.$$

**Theorem:** The continuous RVs  $X_1, \dots, X_n$  are mutually independent if and only if

$$f_{\mathbf{X}}(x_1, \dots, x_n) = f_{X_1}(x_1) \cdots f_{X_n}(x_n).$$

**Proof:** As in the discrete case.

## Examples of Independent Continuous RVs

**1. Minimum of Independent Expo.** Let  $X = \text{Expo}(\lambda)$  and  $Y = \text{Expo}(\mu)$  be independent RVs.

Recall that  $\Pr[X > u] = e^{-\lambda u}$ . Then

$$\begin{aligned}\Pr[\min\{X, Y\} > u] &= \Pr[X > u, Y > u] = \Pr[X > u]\Pr[Y > u] \\ &= e^{-\lambda u} \times e^{-\mu u} = e^{-(\lambda+\mu)u}.\end{aligned}$$

This shows that  $\min\{X, Y\} = \text{Expo}(\lambda + \mu)$ .

Thus, the minimum of two independent exponentially distributed RVs is exponentially distributed.

**2. Minimum of Independent  $U[0, 1]$ .** Let  $X, Y = [0, 1]$  be independent RVs. Let also  $Z = \min\{X, Y\}$ . What is  $f_Z$ ?

One has

$$\Pr[Z > u] = \Pr[X > u]\Pr[Y > u] = (1 - u)^2.$$

Thus  $F_Z(u) = \Pr[Z \leq u] = 1 - (1 - u)^2$ .

Hence,  $f_Z(u) = \frac{d}{du}F_Z(u) = 2(1 - u)$ ,  $u \in [0, 1]$ . In particular,  $E[Z] = \int_0^1 uf_Z(u)du = \int_0^1 2u(1 - u)du = 2\frac{1}{2} - 2\frac{1}{3} = \frac{1}{3}$ .

# Expectation of Product of Independent RVs

**Theorem** If  $X, Y, Z$  are mutually independent, then

$$E[XYZ] = E[X]E[Y]E[Z].$$

**Proof:** Same as discrete case.

**Example:** Let  $X, Y, Z$  be mutually independent and  $U[0, 1]$ . Then

$$\begin{aligned} E[(X + 2Y + 3Z)^2] &= E[X^2 + 4Y^2 + 9Z^2 + 4XY + 6XZ + 12YZ] \\ &= \frac{1}{3} + 4\frac{1}{3} + 9\frac{1}{3} + 4\frac{1}{2}\frac{1}{2} + 6\frac{1}{2}\frac{1}{2} + 12\frac{1}{2}\frac{1}{2} \\ &= \frac{14}{3} + \frac{22}{4} \approx 10.17. \end{aligned}$$



# Variance

**Definition:** The **variance** of a continuous random variable  $X$  is defined as

$$\text{var}[X] = E((X - E(X))^2) = E(X^2) - (E(X))^2.$$

**Example 1:**  $X = U[0, 1]$ . Then

$$\text{var}[X] = E[X^2] - E[X]^2 = \frac{1}{3} - \frac{1}{4} = \frac{1}{12}.$$

**Example 2:**  $X = \text{Expo}(\lambda)$ . Then  $E[X] = \lambda^{-1}$  and  $E[X^2] = 2/(\lambda^2)$ .  
Hence,  $\text{var}[X] = 1/(\lambda^2)$ .

**Example 3:** Let  $X, Y, Z$  be independent. Then

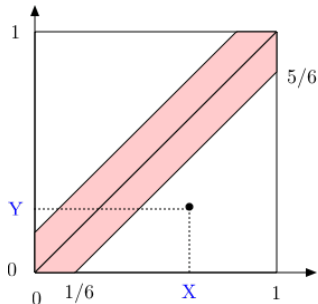
$$\text{var}[X + Y + Z] = \text{var}[X] + \text{var}[Y] + \text{var}[Z],$$

as in the discrete case.

# Meeting at a Restaurant

Two friends go to a restaurant independently uniformly at random between noon and 1pm.

They agree they will wait for 10 minutes. What is the probability they meet?



Here,  $(X, Y)$  are the times when the friends reach the restaurant.

The shaded area are the pairs where  $|X - Y| < 1/6$ , i.e., such that they meet.

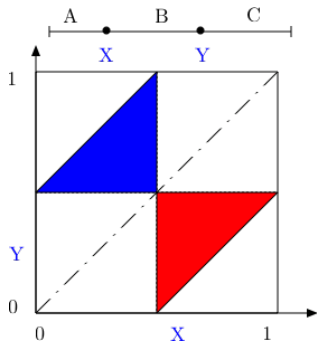
The complement is the sum of two rectangles. When you put them together, they form a square with sides  $5/6$ .

$$\text{Thus, } Pr[\text{meet}] = 1 - \left(\frac{5}{6}\right)^2 = \frac{11}{36}.$$

# Breaking a Stick

You break a stick at two points chosen independently uniformly at random.

What is the probability you can make a triangle with the three pieces?



Let  $X, Y$  be the two break points along the  $[0, 1]$  stick.

You can make a triangle if  $A < B + C$ ,  $B < A + C$ , and  $C < A + B$ .

If  $X < Y$ , this means  $X < 0.5$ ,  $Y < X + 0.5$ ,  $Y > 0.5$ . This is the blue triangle.

If  $X > Y$ , we get the red triangle, by symmetry.

Thus,  $Pr[\text{make triangle}] = 1/4$ .

## Maximum of Two Exponentials

Let  $X = \text{Expo}(\lambda)$  and  $Y = \text{Expo}(\mu)$  be independent. Define  $Z = \max\{X, Y\}$ .

Calculate  $E[Z]$ .

We compute  $f_Z$ , then integrate.

One has

$$\begin{aligned} \Pr[Z < z] &= \Pr[X < z, Y < z] = \Pr[X < z]\Pr[Y < z] \\ &= (1 - e^{-\lambda z})(1 - e^{-\mu z}) = 1 - e^{-\lambda z} - e^{-\mu z} + e^{-(\lambda+\mu)z} \end{aligned}$$

Thus,

$$f_Z(z) = \lambda e^{-\lambda z} + \mu e^{-\mu z} - (\lambda + \mu)e^{-(\lambda+\mu)z}, \forall z > 0.$$

Hence,

$$E[Z] = \int_0^{\infty} z f_Z(z) dz = \frac{1}{\lambda} + \frac{1}{\mu} - \frac{1}{\lambda + \mu}.$$

## Maximum of $n$ i.i.d. Exponentials

Let  $X_1, \dots, X_n$  be i.i.d.  $\text{Expo}(1)$ . Define  $Z = \max\{X_1, X_2, \dots, X_n\}$ .

Calculate  $E[Z]$ .

We use a recursion. The key idea is as follows:

$$Z = \min\{X_1, \dots, X_n\} + V$$

where  $V$  is the maximum of  $n - 1$  i.i.d.  $\text{Expo}(1)$ . This follows from the memoryless property of the exponential.

Let then  $A_n = E[Z]$ . We see that

$$\begin{aligned} A_n &= E[\min\{X_1, \dots, X_n\}] + A_{n-1} \\ &= \frac{1}{n} + A_{n-1} \end{aligned}$$

because the minimum of  $\text{Expo}$  is  $\text{Expo}$  with the sum of the rates.

Hence,

$$E[Z] = A_n = 1 + \frac{1}{2} + \dots + \frac{1}{n} = H(n).$$

# Geometric and Exponential

The geometric and exponential distributions are similar. They are both memoryless.

Consider flipping a coin every  $1/N$  second with  $Pr[H] = p/N$ , where  $N \gg 1$ .

Let  $X$  be the time until the first  $H$ .

**Fact:**  $X \approx \text{Expo}(p)$ .

**Analysis:** Note that

$$\begin{aligned} Pr[X > t] &\approx Pr[\text{first } Nt \text{ flips are tails}] \\ &= \left(1 - \frac{p}{N}\right)^{Nt} \approx \exp\{-pt\}. \end{aligned}$$

Indeed,  $\left(1 - \frac{a}{N}\right)^N \approx \exp\{-a\}$ .

# Summary

## Continuous Probability

- ▶ Continuous RVs are essentially the same as discrete RVs
- ▶ Think that  $X \approx x$  with probability  $f_X(x)\varepsilon$
- ▶ Sums become integrals, ....
- ▶ The exponential distribution is magical: memoryless.