

# Chapter 5 part 1

## Continuous Random Variables

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MATH 241

# Outline

- 1 Continuous random variables
- 2 Expectation and variance of continuous random variable
- 3 Uniform distribution

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# Continuous random variables

👉 A *continuous random variable*  $X$  can take any real value in  $(-\infty, \infty)$ .

## Definition

$X$  is a continuous random variable if there exists a **nonnegative** function  $f$  defined for any  $x \in (-\infty, \infty)$ , such that for any set  $B$  of real numbers,

$$P(X \in B) = \int_B f(x) dx$$

This function  $f$  is called the *probability density function* (pdf) of the random variable  $X$ .

## Examples of continuous random variable

- Rainfall amount for a year.
- Lifetime of your first car.
- Amount of beer consumed on a game day.

# Pdf and cdf

- For pdf to be valid, in addition to being non-negative,

$$\int_{-\infty}^{\infty} f(x)dx = P(-\infty < X < \infty) = 1$$

- Cdf function of continuous random variable

$$F(a) = P(X \leq a) = \int_{-\infty}^a f(x)dx$$

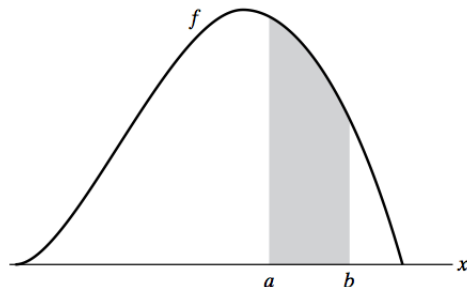
- For continuous random variable, probability to be a single point is zero.

$$P(X = a) = \int_a^a f(x)dx = 0$$

$$P(X < a) = P(X \leq a) - P(X = a) = F(a)$$

- Probability on an interval

$$P(a \leq X \leq b) = F(b) - F(a) = P(a < X < b)$$



$P(a \leq X \leq b) = \text{area of shaded region}$

Connection between pdf and cdf of continuous random variable. If we know the pdf,

$$F(a) = P(X \leq a) = \int_{-\infty}^a f(x)dx$$

How to find the pdf if we know the cdf?

$$f(x) = \frac{d}{dx}F(x)$$

## Question

Suppose that  $X$  is a continuous random variable whose pdf is

$$f(x) = \begin{cases} c(8x - 4x^3) & \text{if } 0 \leq x \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

- Find  $P(X > 0.5)$ .

## Question

Suppose that  $X$  is a continuous random variable whose pdf is

$$f(x) = \begin{cases} c(8x - 4x^3) & \text{if } 0 \leq x \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

- Find  $P(X > 0.5)$ .

$$\begin{aligned} P(X > 0.5) &= \int_{0.5}^{\infty} f(x) dx \\ &= \int_{0.5}^1 \frac{8x - 4x^3}{3} dx \\ &= \frac{1}{3} \left( 4x^2 - x^4 \Big|_{0.5}^1 \right) \\ &= \frac{1}{3} \left[ 3 - 4 \times \left( \frac{1}{2} \right)^2 + \left( \frac{1}{2} \right)^4 \right] = \frac{1}{3} \times \frac{33}{16} = \frac{11}{16} \end{aligned}$$



## Question

Suppose that  $X$  is a continuous random variable whose pdf is

$$f(x) = \begin{cases} c(8x - 4x^3) & \text{if } 0 \leq x \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

- Find the cdf function  $F(x)$ .

$$\int_{-\infty}^{\infty} f(x)dx = 1 \implies c = 1/3$$

$$F(x) = \int_{-\infty}^x f(t)dt = \begin{cases} 0 & x < 0 \\ \frac{4x^2 - x^4}{3} & 0 \leq x \leq 1 \\ 1 & x > 1 \end{cases}$$

Note that  $F(0) = 0$  and  $F(1) = 1$ .

# Interpretation of the pdf

For some small value  $h > 0$ ,

$$\begin{aligned}P\left(x - \frac{h}{2} \leq X \leq x + \frac{h}{2}\right) &= \int_{x-\frac{h}{2}}^{x+\frac{h}{2}} f(t)dt \\&\approx \left[\left(x + \frac{h}{2}\right) - \left(x - \frac{h}{2}\right)\right] f(x) \\&= h \cdot f(x)\end{aligned}$$

The larger  $f(x)$  is, the more likely  $X$  is to be “near”  $x$ .

# Recap

A continuous random variable  $X$  can take more than countable number of values in  $\mathbb{R}$ .

- We defined continuous random variable using pdf

$$P(X \in B) = \int_B f(x)dx$$

- Cdf function of continuous random variable

$$F(x) = P(X \leq x) = \int_{-\infty}^x f(t)dt$$

$$f(x) = \frac{d}{dx}F(x)$$

## Question

Comparing discrete random variables and continuous random variables.

|  | discrete | continuous |
|--|----------|------------|
| values can take                              |          |            |
| pmf / pdf                                    |          |            |
| validness (i.e. the sum of probability is 1) |          |            |
| cdf  |          |            |
| from cdf to pmf / pdf                        |          |            |
| probability at a single point                |          |            |

|  | discrete  | continuous   |
|--|---|--|
| values can take                              | a countable number                                  | any real value in $(-\infty, \infty)$                          |
| pmf / pdf                                    | pmf, $p(x)$   | pdf, $f(x)$  |
| validness (i.e. the sum of probability is 1) | $\sum_{x:p(x)>0} p(x) = 1$                          | $\int_{-\infty}^{\infty} f(x)dx = P(-\infty < X < \infty) = 1$ |
| cdf  | $F(a) = \sum_{x \leq a} p(x)$                       | $F(a) = P(X \leq a) = \int_{-\infty}^a f(x)dx$                 |
| from cdf to pmf / pdf                        | use a graph and pay attention to start / end points | take derivative with respect to the random variable            |
| probability at a single point                | $p(x)$  | 0  |

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# Expectation and variance of continuous random variable

- Recall that expected value of the discrete random variable  $X$  with pmf  $p(x)$ ,

$$E[X] = \sum_{\text{all } x} xp(x)$$

- We define expected value of a continuous random variable  $X$  with pdf  $f(x)$  as

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

- Definitions of variance and standard deviation are the same.

$$Var(X) = E[X^2] - (E[X])^2, \quad SD(X) = \sqrt{Var(X)}$$

Definition:

$$Var(X) = E[(X - \mu)^2]$$

## Properties of $E[X]$ for continuous random variable

- Recall that expected value of the discrete random variable  $X$  with pmf  $p(x)$ ,

$$E[g(X)] = \sum_{\text{all } x} g(x)p(x)$$

- Similarly, expected value of a continuous random variable  $X$  with pdf  $f(x)$  as

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



## Properties of $E[X]$ for continuous random variable

Similarly as discrete random variable, for continuous random variable  $X$  and  $Y$

- Sum of two random variable's

$$E[X + Y] = E[X] + E[Y]$$

- If  $a$  and  $b$  are constants, then

$$E[aX + b] = aE[X] + b$$

- If  $a$  and  $b$  are constants, then

$$Var(aX + b) = a^2 Var(X)$$

## Question

Find  $E[e^X]$ , if the pdf of  $X$  is given by

$$f(x) = \begin{cases} 1 & \text{if } 0 \leq x \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

## Question

Find  $E[e^X]$ , if the pdf of  $X$  is given by

$$f(x) = \begin{cases} 1 & \text{if } 0 \leq x \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

$$E[e^X] = \int_0^1 e^x dx = e - 1$$

## Question

$X$  is a continuous random variable. Its pdf  $f(x)$  is an even function, i.e.,

$$f(-x) = f(x), \quad \text{for any } x > 0$$

What is  $E[X]$ ?

- ☐ a Cannot decide. Need more information.
- ☐ b 0
- ☐ c 1
- ☐ d  $e$

## Question

$X$  is a continuous random variable. Its pdf  $f(x)$  is an even function, i.e.,

$$f(-x) = f(x), \quad \text{for any } x > 0$$

What is  $E[X]$ ?

(a) Cannot decide. Need more information.

(b) 0

(c) 1

(d)  $e$

$$\begin{aligned} E[X] &= \int_{-\infty}^0 x f(x) dx + \int_0^{\infty} x f(x) dx \quad (\text{let } y = -x) \\ &= \int_{-\infty}^0 (-y) f(-y) d(-y) + \int_0^{\infty} x f(x) dx \\ &= \int_{-\infty}^0 y f(-y) dy + \int_0^{\infty} x f(x) dx = - \int_0^{\infty} y f(y) dy + \int_0^{\infty} x f(x) dx \end{aligned}$$

## Question

The density function of  $X$  is given by

$$f(x) = \begin{cases} a + bx^2 & \text{if } 0 \leq x \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

If  $E[X] = \frac{3}{5}$ , find  $a$  and  $b$ .

## Question

The density function of  $X$  is given by

$$f(x) = \begin{cases} a + bx^2 & \text{if } 0 \leq x \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

If  $E[X] = \frac{3}{5}$ , find  $a$  and  $b$ .

$$\begin{aligned} E[X] &= \int_0^1 xf(x)dx = \int_0^1 x(a + bx^2)dx = \int_0^1 ax + bx^3dx \\ &= \frac{1}{2}ax^2 + \frac{1}{4}bx^4 \Big|_0^1 = \frac{a}{2} + \frac{b}{4} = \frac{3}{5} \end{aligned}$$

Moreover,

$$\int_0^1 f(x)dx = \int_0^1 a + bx^2dx = ax + \frac{1}{3}bx^3 \Big|_0^1 = a + \frac{b}{3} = 1$$

Then we have  $a = 3/5$  and  $b = 6/5$ .

## Question

Comparing discrete random variables and continuous random variables.

|                    | discrete | continuous |
|--------------------|----------|------------|
| $E[X]$             |          |            |
| $E[g(X)]$          |          |            |
| $\text{Var}(X)$    |          |            |
| $\sigma$           |          |            |
| $E[aX+b]$          |          |            |
| $\text{Var}(aX+b)$ |          |            |
| $E[X+Y]$           |          |            |



|                    | discrete                   | continuous                           |
|--------------------|----------------------------|--------------------------------------|
| $E[X]$             | $\sum_{x:p(x)>0} xp(x)$    | $\int_{-\infty}^{\infty} xf(x)dx$    |
| $E[g(X)]$          | $\sum_{x:p(x)>0} g(x)p(x)$ | $\int_{-\infty}^{\infty} g(x)f(x)dx$ |
| $\text{Var}(X)$    | $E[X^2] - E[X]^2$          | $E[X^2] - E[X]^2$                    |
| $\sigma$           | $\sqrt{E[X^2] - E[X]^2}$   | $\sqrt{E[X^2] - E[X]^2}$             |
| $E[aX+b]$          | $aE[X] + b$                | $aE[X] + b$                          |
| $\text{Var}(aX+b)$ | $a^2\text{Var}(X)$         | $a^2\text{Var}(X)$                   |
| $E[X+Y]$           | $E[X] + E[Y]$              | $E[X] + E[Y]$                        |

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# Uniform Distribution

Let's define a continuous probability distribution that has some constant value  $c$  between  $\alpha$  and  $\beta$  where  $\alpha < \beta$ . What is the pdf?

$$f(x) = \begin{cases} c & \text{if } \alpha < x < \beta \\ 0 & \text{otherwise} \end{cases}$$

$$1 = \int_{-\infty}^{\infty} f(x)dx = \int_{\alpha}^{\beta} cdx = c(\beta - \alpha) \implies c = \frac{1}{\beta - \alpha}$$

## Definition

A continuous random variable  $X$  has a *Uniform distribution* on the interval  $(\alpha, \beta)$  if its pdf is

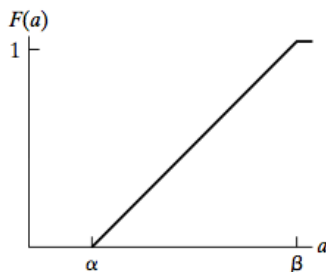
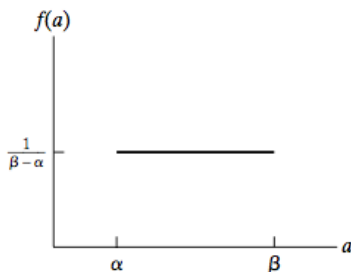
$$X \sim \text{Unif}(\alpha, \beta) \iff f(x) = \frac{1}{\beta - \alpha} \cdot \mathbf{1}_{(\alpha, \beta)}(x) = \begin{cases} \frac{1}{\beta - \alpha} & \text{if } \alpha < x < \beta \\ 0 & \text{otherwise} \end{cases}$$

# Cdf of Uniform distribution

For any  $x \in (\alpha, \beta)$ ,

$$F(x) = \int_{-\infty}^x f(t) dt = \int_{\alpha}^x \frac{1}{\beta - \alpha} dt = \frac{x - \alpha}{\beta - \alpha}$$

- $X \sim \text{Unif}(\alpha, \beta)$ , then its cdf is  $F(x) = \begin{cases} 0 & \text{if } x \leq \alpha \\ \frac{x - \alpha}{\beta - \alpha} & \text{if } \alpha < x < \beta \\ 1 & \text{if } x \geq \beta \end{cases}$



## Mean of a Uniform distribution

- (required) Expected value of  $X \sim \text{Unif}(\alpha, \beta)$  is

$$E[X] = \frac{\alpha + \beta}{2}$$

$$\begin{aligned} E[X] &= \int_{\alpha}^{\beta} \frac{x}{\beta - \alpha} dx = \frac{x^2}{2(\beta - \alpha)} \Big|_{\alpha}^{\beta} \\ &= \frac{\beta^2 - \alpha^2}{2(\beta - \alpha)} = \frac{\alpha + \beta}{2} \end{aligned}$$

# Variance of a Uniform distribution

- (required) Variance of  $X \sim \text{Unif}(\alpha, \beta)$  is

$$\begin{aligned} \text{Var}(X) &= \frac{(\beta - \alpha)^2}{12} \\ E[X^2] &= \int_{\alpha}^{\beta} \frac{x^2}{\beta - \alpha} dx = \frac{x^3}{3(\beta - \alpha)} \Big|_{\alpha}^{\beta} \\ &= \frac{\beta^3 - \alpha^3}{3(\beta - \alpha)} = \frac{\beta^2 + \beta\alpha + \alpha^2}{3} \end{aligned}$$

$$\begin{aligned} \text{Var}(X) &= E[X^2] - E[X]^2 = \frac{\beta^2 + \beta\alpha + \alpha^2}{3} - \frac{(\beta + \alpha)^2}{4} \\ &= \frac{4\beta^2 + 4\beta\alpha + 4\alpha^2}{12} - \frac{3\beta^2 + 6\alpha\beta + 3\alpha^2}{12} \\ &= \frac{\beta^2 - 2\beta\alpha + \alpha^2}{12} = \frac{(\beta - \alpha)^2}{12} \end{aligned}$$

## Uniform distribution: probability calculation

If  $X \sim \text{Unif}(\alpha, \beta)$ , then 
$$P(X \in B) = \frac{\text{length}(B)}{\beta - \alpha}$$

### Question

If  $X$  is uniformly distributed over  $(0, 10)$ , calculate the probability that (a)  $X < 3$ , (b)  $X > 6$ , and (c)  $3 < X < 8$ .

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You can solve this using definitions...

- Part (a):  $P(X < 3) = \int_0^3 \frac{1}{10} dx = \frac{3}{10}$
- Part (b):  $P(X > 6) = \int_6^{10} \frac{1}{10} dx = \frac{4}{10}$
- Part (c):  $P(3 < X < 8) = \int_3^8 \frac{1}{10} dx = \frac{1}{2}$



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- Part (c):  $P(3 < X < 8) = \int_3^8 \frac{1}{10} dx = \frac{1}{2}$

Or recognize the result above to get  $\frac{3}{10}$  directly by  $\frac{3-0}{10}$ . Same for (b) and (c).

# Recap

Expectation for continuous random variable  $X$  and a function of it  $g(X)$

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

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

Uniform distribution  $X \sim \text{Unif}(\alpha, \beta)$

- Pdf

$$f(x) = \begin{cases} \frac{1}{\beta - \alpha} & \text{if } \alpha < x < \beta \\ 0 & \text{otherwise} \end{cases}$$

- Cdf

$$F(x) = \begin{cases} 0 & \text{if } x \leq \alpha \\ \frac{x - \alpha}{\beta - \alpha} & \text{if } \alpha < x < \beta \\ 1 & \text{if } x \geq \beta \end{cases}$$

- Mean and variance

$$E[X] = \frac{\alpha + \beta}{2}, \quad \text{Var}(X) = \frac{(\beta - \alpha)^2}{12}$$