

# Math 461 Fall 2021

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# Outline

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- 1 **General Info**
- 2 5.2 Expectation & Variance of Absolutely Continuous RVs
- 3 5.3 The Uniform Random Variable
- 4 5.4 Normal Random Variables

HW5 is due Friday, 10/01, before the end of class. You can either submit a hard copy or electronically as a pdf file via the HW5 folder in the course Moodle page.

Solutions to HW4 is on my homepage.

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## Theorem

Suppose that  $X$  is an absolutely continuous random variable with density  $f$  and that  $\phi$  is a function on  $\mathbb{R}$ . If

$$\int_{-\infty}^{\infty} |\phi(x)|f(x)dx < \infty,$$

then the random variable  $\phi(X)$  has finite expectation and

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

Suppose  $X$  is an absolutely continuous random variable with finite expectation. For any  $a, b \in \mathbb{R}$ ,

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

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### Example 4

Suppose that  $X$  is an absolutely continuous random variable with density

$$f(x) = \begin{cases} 1, & x \in (0, 1) \\ 0, & \text{otherwise.} \end{cases}$$

Find  $E[e^X]$ .

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

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A stick of length 1 is split at a random point  $U$  with density

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Find the expected length of the piece that contains the point  $p$ ,  $p \in (0, 1)$ .

The length of the piece containing the point  $p$  is

$$L_p(U) = \begin{cases} 1 - U, & U \leq p \\ U, & U > p. \end{cases}$$

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$$\begin{aligned} E[L_p(U)] &= \int_0^1 L_p(u) du = \int_0^p (1-u) du + \int_p^1 u du \\ &= \frac{1}{2} + p(1-p). \end{aligned}$$

### Definition

Suppose that  $X$  is an absolutely continuous random variable with finite expectation  $\mu = E[X]$ . The variance of  $X$  is defined to be

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

One can easily check that

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Let  $f$  be the density of  $X$ . Then

$$\begin{aligned}\text{Var}(X) &= \int_{-\infty}^{\infty} (x - \mu)^2 f(x) dx = \int_{-\infty}^{\infty} (x^2 - 2\mu x + \mu^2) f(x) dx \\ &= \int_{-\infty}^{\infty} x^2 f(x) dx - 2\mu \int_{-\infty}^{\infty} x f(x) dx + \mu^2 \int_{-\infty}^{\infty} f(x) dx \\ &= E[X^2] - 2\mu E[X] + \mu^2 = E[X^2] - (E[X])^2.\end{aligned}$$

Suppose that  $X$  is an absolutely continuous random variable with finite variance, and  $a, b$  are real numbers. Then

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### Example 6

Suppose that  $X$  is an absolutely continuous random variable with density

$$f(x) = \begin{cases} 3x^2, & x \in (0, 1) \\ 0, & \text{otherwise.} \end{cases}$$

Find  $\text{Var}(X)$ .

$$E[X] = \int_0^1 x3x^2 dx = \frac{3}{4}, \quad E[X^2] = \int_0^1 x^2 3x^2 dx = \frac{4}{5}.$$

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## Definition

A random variable  $X$  is said to be uniformly distributed over the interval  $(a, b)$  if its density is given by

$$f(x) = \begin{cases} \frac{1}{b-a}, & x \in (a, b) \\ 0, & \text{otherwise.} \end{cases}$$

If  $X$  is uniformly distributed in  $(a, b)$ , then

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

and

$$E[X^2] = \int_a^b x^2 \frac{1}{b-a} dx = \frac{b^2 + ab + a^2}{3}$$

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### Example 1

Buses arrive at a specified bus stop at 15 minute intervals starting at 7 am. If a passenger arrives at the stop at a time that is uniformly distributed between 7 an 7:30, find the probability that he waits (a) less than 5 minutes; (b) more than 10 minutes.

Let  $X$  be the passenger's arrival time in minutes, after 7 am. Then the answer for (a) is

$$P(10 < X \leq 15) + P(25 < X \leq 30) = \frac{1}{3}.$$

The answer for (b) is

$$P(0 < X \leq 5) + P(15 < X \leq 20) = \frac{1}{3}.$$

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## Example 2

A point is chosen at random on a line segment of length  $L$ . Find the probability that the ratio of the shorter to the longer segment is less than  $\frac{1}{4}$ .

Imagine that the line segment is the interval  $(0, L)$ . Let  $X$  the coordinate of the random chosen point. Then  $X$  is uniformly distributed in  $(0, L)$ . The answer is

$$P\left(\min\left(\frac{X}{L-X}, \frac{L-X}{X}\right) < \frac{1}{4}\right) = 1 - P\left(\frac{L}{5} < X < \frac{4L}{5}\right) = \frac{2}{5}.$$

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Before we introduce the concept of normal random variables, let us look at the function

$$g(x) = e^{-\frac{x^2}{2}}, \quad x \in \mathbb{R}.$$

The function  $g$  is strictly positive, and goes to zero very fast near  $\infty$  and  $-\infty$ , and so

$$c = \int_{-\infty}^{\infty} g(x) dx = \int_{-\infty}^{\infty} e^{-\frac{x^2}{2}} dx$$

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is finite and positive. What is the value of  $c$ ?

$$\begin{aligned}c^2 &= \left( \int_{-\infty}^{\infty} e^{-\frac{x^2}{2}} dx \right) \left( \int_{-\infty}^{\infty} e^{-\frac{y^2}{2}} dy \right) \\&= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} e^{-\frac{x^2+y^2}{2}} dx dy \\&= \int_0^{\infty} \int_0^{2\pi} e^{-\frac{r^2}{2}} r dr d\theta = 2\pi \int_0^{\infty} r e^{-\frac{r^2}{2}} dr = 2\pi.\end{aligned}$$

Thus  $c = \sqrt{2\pi}$  and hence the function

$$\phi(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}, \quad x \in \mathbb{R}$$

is a density function. It is called the standard normal density.

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