CSL003P1M: Probability and Statistics Lecture 29 (Continuous Distributions)

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Introduction

- We have considered discrete random variables whose set of possible values is either finite or countably infinite.
- However, there also exist random variable whose set of possible values is uncountable.
- Some examples (a) a train arrives at a specified stop, (b) the lifetime of a transistor etc.

Continuous Random Variables

Continuous Random Variables

We say that X a continuous random variable if there exists a nonnegative function f, defined for all real $x \in (-\infty, \infty)$, having the property that, for any set B of real numbers,

$$P\{X \in B\} = \int_B f(x)d(x)$$

The function f is called the **probability density function** of the random variable X.

f must satisfy

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

Continuous Random Vaiable

All probability statements about X can be answered in terms of f. For instance,

Let B = [a, b], we obtain

$$P\{a \le X \le b\} = \int_a^b f(x) dx$$

If we let a = b,

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



Continuous Random Vaiable

Cumulative Distribution Function

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

Suppose that X is a continuous random variable whose probability density function is given by

$$f(x) = \begin{cases} C(4x - 2x^2) & 0 < x < 2 \\ 0 & \text{otherwise} \end{cases}$$

- What is the value of C?
- ② Find $P\{X > 1\}$.

Solution: (1) Since f is a probability density function, we must have $\int_{-\infty}^{\infty} f(x)dx = 1$, implying that

$$C\int_{0}^{2} (4x - 2x^{2}) dx = 1$$



$$C\left[2x^2 - \frac{2x^3}{3}\right]_{x=0}^{x=2} = 1$$

Thus,

$$C = 3/8$$
.

$$P{X > 1} = \int_{1}^{\infty} f(x)dx$$
$$= \frac{3}{8} \int_{1}^{2} (4x - 2x^{2})dx$$
$$= \frac{1}{2}$$

The amount of time in hours that a computer functions before breaking down is a continuous random variable with probability density function given by

$$f(x) = \begin{cases} \lambda e^{-x/100} & x \ge 0 \\ 0 & x < 0 \end{cases}$$

What is the probability that

- a computer will function between 50 and 150 hours before breaking down?
- 2 it will function for fewer than 100 hours?

Solution: Since

$$1 = \int_{-\infty}^{\infty} f(x) dx = \lambda \int_{0}^{\infty} e^{-x/100} dx$$

we obtain

$$1 = -\lambda (100)e^{-x/100}|_0^{\infty} = 100\lambda \Rightarrow \lambda = \frac{1}{100}.$$

(1) The probability that a computer will function between 50 and 150 hours before breaking down is

$$P\{50 < X < 150\} = \int_{50}^{150} \frac{1}{100} e^{-x/100} dx = -e^{-x/100}|_{50}^{150}$$
$$= e^{-1/2} - e^{-3/2} \approx 0.384$$

(2)

$$P\{X < 100\} = \int_0^{100} \frac{1}{100} e^{-x/100} dx = -e^{-x/100}|_0^{100} = 1 - e^{-1} \approx 0.633$$

The lifetime in hours of a certain kind of radio tube is a random variable having a probability density function given by

$$f(x) = \begin{cases} 0 & x \le 100 \\ \frac{100}{x^2} & x > 100 \end{cases}$$

What is the probability that exactly 2 of 5 such tubes in a radio set will have to be replaced within the first 150 hours of operation? Assume that the events E_i , i=1,2,3,4,5, that the ith such tube will have to be replaced within this time are independent.

$$P(E_i) = \int_0^{150} f(x) dx$$

$$= 100 \int_{100}^{150} x^{-2} dx$$

$$= \frac{1}{3}$$

Since E_i 's are independent, the probability that exactly 2 of 5 such tubes in a radio set will have to be replaced within the first 150 hours of operation is

$$\binom{5}{2} \left(\frac{1}{3}\right)^2 \left(\frac{2}{3}\right)^3 = \frac{80}{243}$$



Cumulative Distribution Function and Probability Density Function

The relationship between the cumulative distribution F and the probability density f is expressed by

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

Cumulative Distribution Function and Probability Density Function

Differentiating both sides of the preceeding equation yields

$$\frac{d}{da}F(a)=f(a)$$

That is, the density is the derivative of the cumulative distribution function. A somewhat more intuitive interpretation of the density function may be obtained as follows:

$$P\left\{a-\frac{\epsilon}{2}\leq X\leq a+\frac{\epsilon}{2}\right\}=\int_{a-\frac{\epsilon}{2}}^{a+\frac{\epsilon}{2}}f(x)dx\approx \epsilon f(a)$$



If X is continuous with distribution function F_X and density function f_X , find the density function of Y = 2X.

Solution: First approach

$$F_Y(a) = P\{Y \le a\}$$

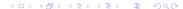
$$= P\{2X \le a\}$$

$$= P\{X \le a/2\}$$

$$= F_X(a/2)$$

Differentiation gives

$$f_Y(a) = \frac{1}{2} f_X(a/2)$$



Second approach:

$$\epsilon f_Y(a) = P\{a - \frac{\epsilon}{2} \le Y \le a + \frac{\epsilon}{2}\}
= P\{a - \frac{\epsilon}{2} \le 2X \le a + \frac{\epsilon}{2}\}
= P\{\frac{a}{2} - \frac{\epsilon}{4} \le X \le \frac{a}{2} + \frac{\epsilon}{4}\}
\approx \frac{\epsilon}{2} f_X(a/2)$$

Dividing through by ϵ gives the same results as before.



Expectation of Continuous Random Variables

We defined the expected value of a discrete random variable X by

$$E[X] = \sum_{x} x P\{X = x\}$$

If X is a continuous random variable having probability density function f(x), then, because

$$f(x)dx \approx P\{x \le X \le x + dx\}$$
 for dx small

it is easy to see that the analogous definition is to define the expected value of X by

Expected Value of a Continuous Random Variable X

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



Find E[X] when the density function of X is

$$f(x) = \begin{cases} 2x & \text{if } 0 \le x \le 1\\ 0 & \text{otherwise} \end{cases}$$

Solution:

$$E[X] = \int_{-\infty}^{\infty} xf(x)dx$$
$$= \int_{0}^{1} 2x^{2}dx$$
$$= \frac{2}{3}$$

The density function of X is given by

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

Find $E[e^X]$.

Solution: Let $Y = e^X$. We first determine F_Y , the probability density function of Y. Now, for $1 \le x \le e$,

$$F_{Y}(x) = P\{Y \le x\}$$

$$= P\{e^{X} \le x\}$$

$$= P\{X \le \log(x)\}$$

$$= \int_{0}^{\log(x)} f(t)dt$$

$$= \log(x)$$

By differentiating $F_Y(x)$, we get

$$f_Y(x) = \frac{1}{x}$$
 $1 \le x \le e$

Hence,

$$E[e^{X}] = E[Y] = \int_{-\infty}^{\infty} x f_{Y}(x) dx$$
$$= \int_{1}^{e} dx$$
$$= e - 1$$

Thank You