Lecture 9: Sept 12

Last time

• Presentations

Today

- HW2 posted (due: Sept 15th)
- No lecture, but reviews on Fridays with two presentations
- Random variables
- Distribution Functions
- Types of Random Variables

Some properties of the cdf:

Let F(x) be a cdf. Then

1.
$$0 \le F(x) \le 1$$

2.
$$\lim_{x \to -\infty} F(x) = 0$$

$$3. \lim_{x \to \infty} F(x) = 1$$

- 4. F is nondecreasing: if a < b, then $F(a) \leq F(b)$
- 5. F is right-continuous: $\lim_{x \downarrow b} F(x) = F(b)$, or $\lim_{x \to b^+} F(x) = F(b)$
- 6. $Pr(a < X \le B) = F(b) F(a)$

Theorem The function F(x) is a cdf if and only if the following three conditions hold:

- 1. $\lim_{x \to -\infty} F(x) = 0$ and $\lim_{x \to \infty} F(x) = 1$
- 2. F is nondecreasing: if a < b, then $F(a) \leq F(b)$
- 3. F is right-continuous: $\lim_{x\downarrow b} F(x) = F(b)$, or $\lim_{x\to b^+} F(x) = F(b)$

The cdf does not contain information about the original sample space.

Definition Two random variables X and Y are identically distributed if, for every Borel set $A \subset \mathbb{R}$, $\Pr(X \in A) = \Pr(Y \in A)$.

Example Toss a fair coin n times. The number of heads and the number of tails have the same distribution.

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Theorem The following two statements are equivalent:

- 1. The random variables X and Y are identically distributed.
- 2. $F_X(x) = F_Y(x)$ for every x.

Types of Random Variables

Definition A random variable X can be

- discrete:
 - X takes on a finite or countably infinite number of values
 - $-F_X(x)$ is step-wise constant
- continuous:
 - the range of X consists of subsets of the real line
 - $-F_X(x)$ is continuous.
- mixed: $F_X(x)$ is piecewise continuous.

Example A random variable has cdf

$$F(x) = \begin{cases} 0 & x < 0 \\ x/2 & 0 \le x < 1 \\ 2/3 & 1 \le x < 2 \\ 11/12 & 2 \le x < 3 \\ 1 & 3 \le x \end{cases}$$

Is this a valid cdf? Is it a discrete random variable or continuous random variable or mixed? solution:

Discrete Random Variables

Suppose a random variable X takes only a finite or countable number of values. Let the sample space of X be $S = \{x_1, x_2, \dots\}$. Then the cdf can be expressed as:

$$F(x) = \sum_{x_i \leqslant x} \Pr(X = x_i).$$

Definition The probability mass function (pmf) of a discrete random variable X is given by

$$f_X(x) = \Pr(X = x)$$
 for all x .

If the sample space of X is $X = \{x_1, x_2, \dots\}$, then

$$f(x_i) = \Pr(X = x_i) = \Pr(x_{i-1} < X \le x_i) = F(x_i) - F(x_{i-1}).$$

Example (Geometric probabilities) Suppose we do an experiment that consists of tossing a coin until a head appears. Let p = probability of a head on any given toss, and define a random variable X = number of tosses required to get a head. Then for any $x = 1, 2, \ldots$,

$$\Pr(X = x) = (1 - p)^{x - 1} p,$$

since we must get x-1 tails followed by a head for the event to occur and all trials are independent. What is the pmf of the above Geometric distribution? What is the cdf?

solution:

Definition The domain of a random variable X is the set of all values of x for which f(x) > 0. This is also called range or sample space.

Properties of the pmf:

- 1. f(x) > 0 for at most a countable number of values x. For all other values x, f(x) = 0.
- 2. Let $\{x_1, x_2, \dots\}$ denote the domain of X. Then

$$\sum_{i=1}^{\infty} f(x_i) = 1.$$

An obvious consequence is that $f(x) \leq 1$ over the domain.

Example What is the pmf of a deterministic random variable (a constant)? *solution:*

Example In many applications, a formula can be used to represent the pmf of a random variable. Suppose X can take values $1, 2, \ldots$ with pmf

$$f(x) = \begin{cases} \frac{1}{x(x+1)} & \text{for } x = 1, 2, \dots \\ 0 & \text{otherwise.} \end{cases}$$

How would we determine if this is an allowable pmf? solution:

Continuous Random Variables

Definition A random variable X is continuous if $F_X(x)$ is a continuous function of x.

Definition A random variable X is absolutely continuous if $F_X(x)$ is an absolutely continuous function of x.

Definition A function F(x) is absolutely continuous if it can be written

$$F(x) = \int_{-\infty}^{x} f(x)dx.$$

Absolute continuity is stronger than continuity but weaker than differentiability. An example of an absolutely continuous function is one that is:

- continuous everywhere
- differentiable everywhere, except possibly for a countable number of points.

Definition The probability density function or pdf, $f_X(x)$, of a continuous random variable X is the function that satisfies

$$F_X(x) = \int_{-\infty}^x f_X(t)dt$$
 for all x .

Notation: We write $X \sim F_X(x)$ for the expression "X has a distribution given by $F_X(x)$ " where we read the symbol " \sim " as "is distributed as". Similarly, we can write $X \sim f_X(x)$ or , if X and Y have the same distribution, $X \sim Y$.

Theorem A function $f_X(x)$ is a pdf (or pmf) of a random variable X if and only if

- 1. $f_X(x) \ge 0$ for all x.
- 2. $\int_{-\infty}^{\infty} f_X(x) dx = 1$ (pdf) or $\sum_x f_X(x) = 1$ (pmf).

Example Suppose $F(x) = 1 - e^{-\lambda x}$ for x > 0 and F(x) = 0 otherwise. Is F(x) a cdf? What is the associated pdf? solution:

Notes

- If X is a continuous random variable, then f(x) is not the probability that X = x. In fact, if X is an absolutely continuous random variable with density function f(x), then Pr(X = x) = 0. (Why?)

 proof
- Because Pr(X = a) = 0, all the following are equivalent:

$$\Pr(a \leqslant X \leqslant b), \quad \Pr(a \leqslant X < b) \quad , \quad \Pr(a < X \leqslant b) \quad \text{and} \quad \Pr(a < X < b)$$

• f(x) can exceed one!