## Recall

### Cumulative distribution function

The *cumulative distribution function* (CDF) of a random variable X is defined by

$$F(t) \coloneqq P\{X \le t\}, \quad \forall t \in \mathbb{R}$$

which has the following properties:

• Non-decreasing. • Right-continuous. •  $\lim_{t\to-\infty} F(t) = 0$  and  $\lim_{t\to+\infty} F(t) = 1$ .

For  $x \in \mathbb{R}$ ,  $P\{X < x\} = \lim_{t \to x^-} F(t)$ .

#### Continuous random variable

A random variable X is (absolutely) continuous if there exists a non-negative function, called probability density function (PDF), such that

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

where B is a 'measurable' set in  $\mathbb{R}$ . Fortunately, countable unions and intersections of intervals are 'measurable'.

Below are some facts about a **continuous** random variable X:

Unit integral of **a** PDF.  $\int_{-\infty}^{+\infty} f(x) dx = 1$ .

Zero probability at any point.  $\forall x \in \mathbb{R}, P\{X = x\} = 0.$ 

Cumulative distribution function.  $\forall t \in \mathbb{R}, F(t) \coloneqq \int_{-\infty}^{t} f(x) dx.$ 

For  $t \in \mathbb{R}$ , it follows from  $F(t) = P\{X \leq t\} = P\{X < t\} = \lim_{x \to t^-} F(x)$  that F(t) is leftcontinuous, hence continuous, at t. In conclusion, the CDF of a continuous r.v. is continuous.

Expectation.  $E[X] \coloneqq \int_{-\infty}^{+\infty} x f(x) \, dx.$ 

Continuous layer-cake. If X is continuous and non-negative, then  $E[X] = \int_0^{+\infty} P\{X > t\} dt$ .

Expectation of a function of a continuous r.v.. Let  $g: \mathbb{R} \to \mathbb{R}$ . Then  $E[g(X)] = \int_{-\infty}^{+\infty} g(x)f(x) dx$ .

Variance. 
$$\operatorname{Var}(X) := E[(X - E[X])^2] = E[X^2] - (E[X])^2.$$

Affine transform. For  $a, b \in \mathbb{R}$ ,  $\begin{cases} E[aX+b] = aE[X] + b;\\ \operatorname{Var}(aX+b) = a^2 \operatorname{Var}(X). \end{cases}$ 

Relation between PDF f and CDF F. If f is continuous at  $x \in \mathbb{R}$ , then  $F'(x) = \frac{dF(x)}{dx} = f(x)$ .

# Probability computation from CDF

**Example 1.** Suppose a random variable X has CDF

$$F(t) = \begin{cases} 0 & t \in (-\infty, 0) \\ t/4 & t \in [0, 1) \\ 1/2 + (t-1)/4 & t \in [1, 2) \\ 11/12 & t \in [2, 3) \\ 1 & t \in [3, +\infty). \end{cases}$$

Find  $P\{X = i\}$ , i = 1, 2, 3 and  $P\{1 \le X < 3\}$ .

Solution. Below is the graph of F(t).



Then

$$P\{X=1\} = P\{X \le 1\} - P\{X < 1\} = F(1) - \lim_{t \to 1^{-}} F(t) = \frac{1}{2} - \frac{1}{4} = \frac{1}{4},$$
  

$$P\{X=2\} = P\{X \le 2\} - P\{X < 2\} = F(2) - \lim_{t \to 2^{-}} F(t) = \frac{11}{12} - \frac{3}{4} = \frac{1}{6},$$
  

$$P\{X=3\} = P\{X \le 3\} - P\{X < 3\} = F(3) - \lim_{t \to 3^{-}} F(t) = 1 - \frac{11}{12} = \frac{1}{12}$$

And

$$P\{1 \le X < 3\} = P\{X < 3\} - P\{X < 1\} = \lim_{t \to 3^{-}} F(t) - \lim_{t \to 1^{-}} F(t) = \frac{11}{12} - \frac{1}{4} = \frac{2}{3}.$$

*Remark.* Since the CDF of a discrete random variable should be like a step function, it follows that X in Example 1 is not discrete. On the other hand, X is not a continuous random variable either because the CDF of a continuous random variable should be continuous.

## Some computations about continuous random variables

**Example 2.** Let X be a random variable with PDF

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

Find the value of c and the CDF of X.

Solution. Since f is a PDF, we have

$$1 = \int_{-\infty}^{\infty} f(x) \, dx = \int_{-1}^{1} c(1-x^2) \, dx = c(x-\frac{x^3}{3}) \Big|_{-1}^{1} = \frac{4}{3}c,$$

which implies  $c = \frac{3}{4}$ . Recall that for  $t \in \mathbb{R}$ , the CDF  $F(t) \coloneqq \int_{-\infty}^{t} f(x) dx$ .

If 
$$t \leq -1$$
, then  $F(t) = \int_{-\infty}^{t} f(x)dx = \int_{-\infty}^{t} 0dx = 0$ ,  
If  $1 < t \leq 1$ , then  $F(t) = \int_{-\infty}^{t} f(x)dx = \int_{-1}^{t} \frac{3}{4}(1-x^{2})dx = \frac{3}{4}(t-\frac{t^{3}}{3}+\frac{2}{3}) = -\frac{t^{3}}{4} + \frac{3t}{4} + \frac{1}{2}$ ,  
If  $t > 1$ , then  $F(t) = P(X \leq t) = 1 - P(X > t) = 1 - \int_{t}^{\infty} 0dx = 1$ .

Thus

$$F(t) = \begin{cases} 0 & t \in (-\infty, -1] \\ -\frac{t^3}{4} + \frac{3t}{4} + \frac{1}{2} & t \in (-1, 1] \\ 1 & t \in (1, \infty). \end{cases}$$

**Example 3.** Let X be a random variable with PDF  $f_X$ . Find a PDF of random variable Y = aX + b where  $0 \neq a \in \mathbb{R}, b \in \mathbb{R}$ .

Solution. Let  $F_X$  and  $F_Y$  denote the CDFs of X and Y respectively. For  $t \in \mathbb{R}$ ,

$$F_Y(t) = P\{Y \le t\} = P\{aX + b \le t\}.$$

If a > 0, then  $F_Y(t) = P\{X \le \frac{t-b}{a}\} = F_X(\frac{t-b}{a})$ . When  $F_X$  is differentiable at  $\frac{t-b}{a}$ , by chain rule,

$$f_Y(t) = \frac{dF_Y(t)}{dt} = \frac{1}{a}f_X(\frac{t-b}{a}).$$

When  $F_X$  is NOT differentiable at  $\frac{t-b}{a}$ , we define  $f_Y(t) = \frac{1}{a} f_X(\frac{t-b}{a})$ . Together, when a > 0, a possible PDF of Y is

$$f_Y(t) = \frac{1}{a} f_X(\frac{t-b}{a}) \quad , \forall t \in \mathbb{R}.$$

If a < 0, then  $F_Y(t) = P\{X \ge \frac{t-b}{a}\} = 1 - P\{X < \frac{t-b}{a}\} = 1 - P\{X \le \frac{t-b}{a}\} = 1 - F_X(\frac{t-b}{a})$ . We omit the discussion about differentiability. By differentiation, when a < 0, a PDF of Y is

$$f_Y(t) = \frac{dF_Y(t)}{dt} = -\frac{1}{a}f_X(\frac{t-b}{a}) \quad , \forall t \in \mathbb{R}.$$

*Remark.* In Example 3, we have carefully dealt with the differentiability of a CDF in the case of a > 0, which is the rigorous way to think about it. However, in practice we **omit** the discussion because we know that a CDF is differentiable at **most** points. Then as in Example 3, we adjust the values on the **tiny** part of non-differentiable points to simplify the final results.

Let f be a PDF of a continuous random variable X. After changing values of f on a **tiny** part of  $\mathbb{R}$ , the resulted f is still a PDF of X.

*Remark.* Let X be a continuous random variable and  $g: \mathbb{R} \to \mathbb{R}$  be any function. The following example shows that we are not even sure whether g(X) has a PDF. Actually, in Example 3 we have **omitted** the step to prove that Y = aX + b is indeed continuous with a PDF. In practice, when the question asks for a PDF, we can take it for granted that the target PDF exists like Example 3 and Example 4.

**Example 4.** Suppose the CDF of X is

$$F(t) = \begin{cases} 1 - e^{-t^2} & t > 0\\ 0 & t \le 0. \end{cases}$$

Find  $P\{X > 2\}$  and a PDF of X.

Solution. First

$$P\{X > 2\} = 1 - P\{X \le 2\} = 1 - F(2) = e^{-4}.$$

Then

If 
$$x > 0$$
, then  $\frac{dF(x)}{dx} = 2xe^{-x^2}$ .  
If  $x < 0$ , then  $\frac{dF(x)}{dx} = 0$ .

Define

$$f(x) = \begin{cases} 2xe^{-x^2} & x > 0, \\ 0 & x \le 0. \end{cases}$$

Hence f(x) is a PDF of X.