

# Continuous Random Variables

## Definition: Continuous Random Variable

To say that a random variable  $X$  is *continuous* means that its range is an interval or a union of intervals.

## Definition: Probability Density Function

The *Probability Density Function* (pdf) of a continuous random variable  $X$  is a function  $f : \mathbb{R} \rightarrow \mathbb{R}$  such that:

1.  $\forall x \in \mathbb{R}, f(x) \geq 0$
2.  $\int_{-\infty}^{\infty} f(x)dx = 1$
3.  $\forall a, b \in \mathbb{R}$  such that  $a \leq b$ :

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

## Properties: Probability Density Function

Let  $X$  be a continuous random variable:

- $\text{Range}(X) = \{x \mid f(x) > 0\}$
- $x \notin \text{Range}(X) \iff f(x) = 0$
- $P(X = c) = \int_c^c f(x)dx = 0$

Since the probability of a single value is 0, the endpoints of an interval do not matter:

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

## Definition: Cumulative Distribution Function

The *Cumulative Distribution Function* (cdf) of a continuous random variable  $X$  is a function  $F : \mathbb{R} \rightarrow \mathbb{R}$  such that:

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

The *complementary cdf*, denoted  $\bar{F}(x)$ , is given by:

$$\bar{F}(x) = 1 - F(x)$$

## Properties: Cumulative Distribution Function

Let  $X$  be a continuous random variable with pdf  $f(x)$  and cdf  $F(x)$ :

- $\lim_{x \rightarrow -\infty} F(x) = 0$

- $\lim_{x \rightarrow \infty} F(x) = 1$
- $F(x)$  is non-decreasing and continuous on  $\mathbb{R}$ .
- $P(a < X) = 1 - F(a)$
- $P(a < X < b) = F(b) - F(a)$
- $F'(x) = f(x)$

### **Definition: Median**

The *median* of a continuous random variable  $X$  with pdf  $f(x)$  and cdf  $F(x)$ , denoted  $\tilde{\mu}$ , is given by:

$$\frac{1}{2} = F(\tilde{\mu}) = \int_{-\infty}^{\tilde{\mu}} f(x)dx$$

### **Definition: Expected Value**

The *expected value (mean)* of a continuous random variable  $X$  with pdf  $f(x)$ , denoted  $\mu_X$  or  $E(X)$ , is given by:

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

Similarly, for some function  $h(x)$ :

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

### **Definition: Variance**

The *variance* of a continuous random variable  $X$  with pdf  $f(x)$  and expected value  $\mu$ , denoted  $\sigma^2$  or  $V(X)$ , is given by:

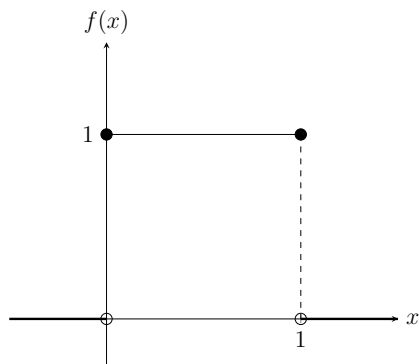
$$\sigma^2 = V(x) = E((X - \mu)^2) = \int_{-\infty}^{\infty} (x - \mu)^2 f(x)dx = E(X^2) - \mu^2 = E(X^2) - E(X)^2$$

The *standard deviation*, denoted  $\sigma$ , is given by:

$$\sigma = \sqrt{\sigma^2} = \sqrt{V(X)}$$

### **Example**

Let  $f(x) = 1, 0 \leq x \leq 1$  be a pdf for a continuous random variable  $X$ :

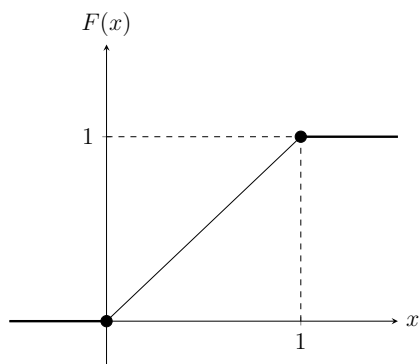


Since this pdf is simple, we can use area to calculate probabilities:

1.  $P(X < -1) = 0$
2.  $P(X = 0.2) = 0$
3.  $P(X < 0.2) = 0.2 \cdot 1 = 0.2$
4.  $P(0.2 < X < 0.5) = (0.5 - 0.2) \cdot 1 = 0.3$
5.  $P(X > 0.6) = (1 - 0.6) \cdot 1 = 0.5$

To find the cdf:

$$F(x) = \int_0^x dt = t \Big|_0^x = x$$



To find the median:

$$\frac{1}{2} = F(\tilde{\mu}) = \tilde{\mu}$$

To find the expected value:

$$E(X) = \int_0^1 x \cdot 1 dx = \int_0^1 x dx = \frac{1}{2} x^2 \Big|_0^1 = \frac{1}{2}$$

To find the variance and standard deviation:

$$E(X^2) = \int_0^1 x^2 \cdot 1 dx = \int_0^1 x^2 dx = \frac{1}{3} x^3 \Big|_0^1 = \frac{1}{3}$$

$$\sigma^2 = \frac{1}{3} - \left(\frac{1}{2}\right)^2 = \frac{1}{3} - \frac{1}{4} = \frac{1}{12}$$

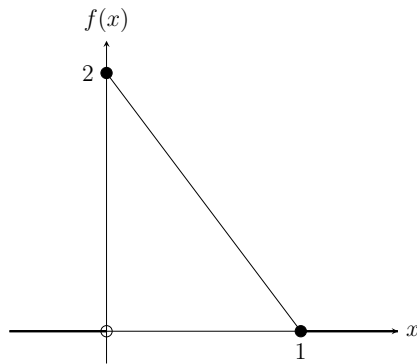
$$\sigma = \sqrt{\frac{1}{12}} = \frac{1}{2\sqrt{3}} = \frac{1}{6}\sqrt{3}$$

### Example

Let  $f(x) = c(1-x)$ ,  $0 < x < 1$  be the pdf for a continuous random variable  $X$ . We first need to find an appropriate value for  $c$ :

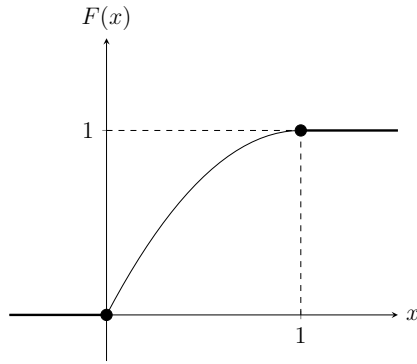
$$\int_0^1 c(1-x) dx = c \left( x - \frac{1}{2}x^2 \right) \Big|_0^1 = \frac{1}{2}c = 1$$
$$c = 2$$

And so  $f(x) = 2(1-x)$ :



To find the cdf:

$$F(x) = \int_0^x 2(1-t) dt = \int_0^x (2-2t) dt = (2t - t^2) \Big|_0^x = 2x - x^2 = x(2-x)$$



To find the median:

$$\frac{1}{2} = F(\tilde{\mu}) = 2\tilde{\mu} - \tilde{\mu}^2$$

$$\tilde{\mu}^2 - 2\tilde{\mu} + \frac{1}{2} = 0$$

$$2\tilde{\mu}^2 - 4\tilde{\mu} + 1 = 0$$

$$\tilde{\mu} = \frac{4 \pm \sqrt{16 - 8}}{4} = 1 \pm \frac{1}{2}\sqrt{2} = 1 \pm \frac{1}{\sqrt{2}}$$

Honoring the domain:

$$\tilde{\mu} = 1 - \frac{1}{\sqrt{2}} = 0.2929$$

To find the expected value:

$$E(X) = \int_0^1 x \cdot 2(1-x)dx = \int_0^1 (2x - 2x^2)dx = \left(x^2 - \frac{2}{3}x^3\right)\Big|_0^1 = 1 - \frac{2}{3} = \frac{1}{3}$$

To find the expected value:

$$E(X) = \int_0^1 x \cdot 2(1-x)dx = \int_0^1 (2x - 2x^2)dx = \left(x^2 - \frac{2}{3}x^3\right)\Big|_0^1 = 1 - \frac{2}{3} = \frac{1}{3}$$

To find the variance and standard deviation:

$$E(X^2) = \int_0^1 x^2 \cdot 2(1-x)dx = \int_0^1 (2x^2 - 2x^3)dx = \left(\frac{2}{3}x^3 - \frac{1}{2}x^4\right)\Big|_0^1 = \frac{2}{3} - \frac{1}{2} = \frac{1}{6}$$

$$\sigma^2 = \frac{1}{6} - \left(\frac{1}{3}\right)^2 = \frac{1}{6} - \frac{1}{9} = \frac{1}{18}$$

$$\sigma = \sqrt{\frac{1}{18}} = \frac{1}{3\sqrt{2}} = \frac{1}{6}\sqrt{2}$$