

## § 3.4 Probability Density Functions

The defn of probability density fns for continuous r.v.'s is similar to that for discrete r.v.'s, except for a couple of 'surprises'!

Defn 3.4 A fn  $f: \mathbb{R} \rightarrow \mathbb{R}$  is called a probability density function of the continuous random variable  $X$  iff

$$P([a, b]) = \int_a^b f(x) dx$$

Often called densities or just PDFs.

Note that  $f(c)$  does not give  $P(X=c)$  as it does in the discrete case.

For cts r.v.'s we always have

$$P(X=c) = 0 \quad \text{for any } c \in \mathbb{R}.$$

(+g. the highway example).

Because of this, we can change 'some' of the values of  $f(x)$  and still have a PDF for  $X$ . Hence, the PDF of  $X$  is not uniquely defined (as it is in the discrete case).

Similarly, it doesn't matter from the point of view of probability whether or not we include the endpoints of an interval.

Thm 3.4 If  $X$  is a cts r.v. and  $a, b \in \mathbb{R}$ , then

$$P([a, b]) = P([a, b)) = P((a, b]) = P((a, b)).$$

We also have an analogue of Thm 3.1 from the last section.

Thm 3.5 A p.d.f. can serve as the PDF of a cts r.v.  $X$  if its values,  $f(x)$ , satisfy

1.  $f(x) \geq 0$ ,  $-\infty < x < \infty$

2.  $\int_{-\infty}^{\infty} f(x) dx = 1.$

Ex. If  $X$  has the density

$$f(x) = \begin{cases} ke^{-3x}, & x > 0 \\ 0, & x \leq 0 \end{cases}$$

find  $k$  and  $P([0.5, 1])$ .

Soln. By Thm 3.5, must have

$$\int_{-\infty}^{\infty} f(x) dx = 1.$$

$$\text{Now } \int_{-\infty}^{\infty} f(x) dx = \int_0^{\infty} k e^{-3x} dx$$

$$= k \lim_{t \rightarrow \infty} \int_0^t e^{-3x} dx$$

$$= k \lim_{t \rightarrow \infty} \int_0^t e^{-3x} dx$$

$$= k \lim_{t \rightarrow \infty} \left[ \frac{e^{-3x}}{-3} \right]_0^t$$

$$= k \lim_{t \rightarrow \infty} \left[ \frac{e^{-3t}}{-3} - \left( -\frac{e^0}{3} \right) \right]$$

$$= k \lim_{t \rightarrow \infty} \left( \frac{1}{3} - \frac{e^{-3t}}{3} \right)$$

$$= \frac{k}{2}$$

Hence in order to have  $\int_{-\infty}^{\infty} f(x) dx = 1$ , we need

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

$$\text{or } k = 3.$$

We then get

$$P([0.5, 1]) = \int_{0.5}^1 3e^{-3x} dx = \left[ -e^{-3x} \right]_{0.5}^1$$

$$= -e^{-3} + e^{-0.5} \approx 0.173.$$

Actually, the r.v. in this example cannot assume negative values, but we extended its PDF to all of  $\mathbb{R}$ .

As in the discrete case, we are often interested in  $P(X \leq x)$ .

Defn 3.5 If  $X$  is a cts r.v. and the value of a PDF for  $X$  at  $t$  is  $f(t)$ , then the  $f_0$  given by

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

is called the cumulative distribution function (CDF) of  $X$ .

Note that although  $f(t)$  is not uniquely defined,  $F(x)$  is as  $P(X \leq x)$  is uniquely defined!

As with the discrete case we can say that

$$\lim_{x \rightarrow -\infty} F(x) = 0, \quad \lim_{x \rightarrow \infty} F(x) = 1 \quad \text{and}$$

$$a, b \in \mathbb{R}, \quad a < b \Rightarrow F(a) \leq F(b)$$

(Almost) Formally, a continuous random variable is one whose CDF is continuous.

Also have

Thm 3.6 If  $f(x)$ ,  $F(x)$  are a PDF  
and the CDF for a r.v.  $X$ , then

$$P([a, b]) = F(b) - F(a), \quad a, b \in \mathbb{R}, \quad a \leq b$$

and

$$f(x) = \frac{dF(x)}{dx}$$

where this derivative exists.

PF. (Not fully rigorous)

For first part

$$P((-\infty, b]) = P((-\infty, a) \cup [a, b]) \quad (\text{disjoint})$$

$$\begin{aligned} \text{So } P((-\infty, b]) &= P((-\infty, a) \cup [a, b]) \\ &= P((-\infty, a]) + P([a, b]) \quad \text{by Thm 3.4} \end{aligned}$$

$$\text{ü } F(b) = F(a) + P([a, b]), \quad \text{so}$$

$$P([a, b]) = F(b) - F(a) \quad \text{as desired.}$$

For second part, assume  $f$  is cb. at  $x$ .

$\Rightarrow$  if  $t$  is near  $x$ , then  $f(t) \approx f(x)$ .

Now suppose  $h$  is small (true or -ve).  
( $\neq 0$ )

$$\frac{F(x+h) - F(x)}{h} = \frac{1}{h} \left( \int_{-\infty}^{x+h} f(t) dt - \int_{-\infty}^x f(t) dt \right)$$

$$= \frac{1}{h} \int_x^{x+h} f(t) dt$$

$$\approx \frac{1}{h} \cdot h f(x)$$

$$= f(x).$$

(Mean Value Theorem)

Hence if  $f$  is cb. at  $x$ ,

$\lim_{h \rightarrow 0} \frac{F(x+h) - F(x)}{h}$  exists (i.e.  $F$  is differentiable at  $x$ )

and  $\frac{dF(x)}{dx} = f(x)$ .



Ex. Find the CDF of the r.v.  $X$  of the last example and use it to reevaluate  $P([0.5, 1])$ .

Soln. For  $x > 0$

$$F(x) = \int_{-\infty}^x f(t) dt = \int_0^x 3e^{-3t} dt = [-e^{-3t}]_0^x = 1 - e^{-3x}$$

and since  $F(x) = 0$  for  $x \leq 0$ , we can write

$$F(x) = \begin{cases} 0, & x \leq 0 \\ 1 - e^{-3x}, & x > 0 \end{cases}$$

By Thm 3.9,  $P([0.5, 1]) = F(1) - F(0.5)$

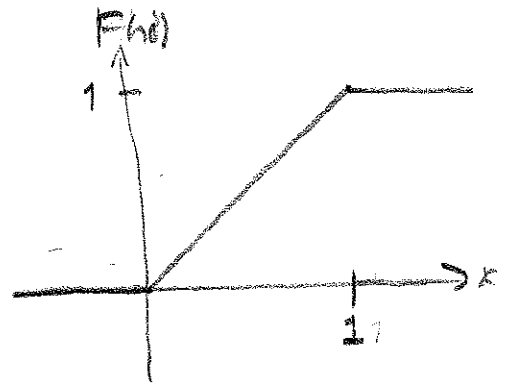
$$\begin{aligned} &= (1 - e^{-3}) - (1 - e^{-1.5}) \\ &= e^{-1.5} - e^{-3} \\ &= 0.173 \end{aligned}$$

(same answer as before). □



Ex. Find a PDF for the r.v. whose CDF is

$$F(x) = \begin{cases} 0, & x \leq 0 \\ x, & 0 < x < 1 \\ 1, & x \geq 1 \end{cases}$$



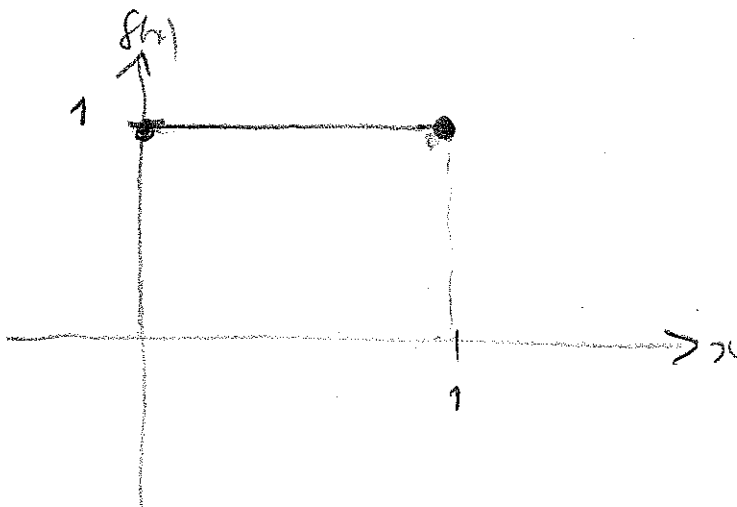
Soln.  $F$  is diff everywhere except at  $0, 1$ .

By Thm 3.9, we have

$$f(x) = \begin{cases} 0, & x < 0 \\ 1, & 0 < x < 1 \\ 0, & x > 1 \end{cases}$$

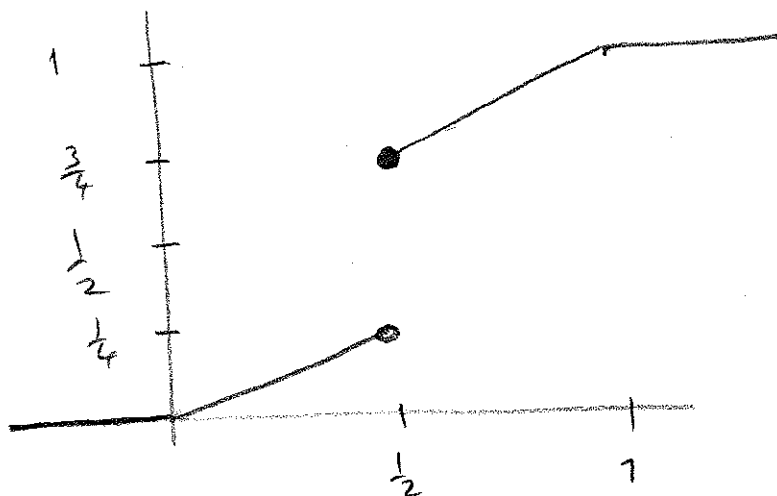
Still need to define  $f$  at  $0, 1$ . However, as we saw earlier, the particular values of  $f$  at  $0, 1$  will not affect the probs. of events.

So set  $f(0) = f(1) = 1$ .



Also have r.v.s which are neither discrete nor cts.

e.g. Consider a r.v. with CDF



This r.v. takes on uncountably many values so it is not discrete.

However,  $P(\{\frac{1}{2}\}) = \frac{1}{2} > 0$ , so it is not ct either (prob. of  $\{\frac{1}{2}\}$  is the height of the jump in the CDF at  $\frac{1}{2}$ ).

In our course, we shall limit ourselves to r.v.s which are discrete or cts with the CDFs of the cts. r.v.s being piecewise differentiable (i.e. differentiable on all of  $\mathbb{R}$  except (possibly) at finitely many values).