CONTINUOUS RANDOM VARIABLES (Chapter 4)

Cumulative distribution functions

The (cumulative) distribution function (cdf) of a random variable Y is

$$F(y) = P(Y \le y)$$
, $y \in \mathbb{R}$

Example 1 Let Y be the number of heads that come up on 2 tosses of a coin.

Find *Y*'s cdf.

Find
$$Y \le cdf$$
.

$$R(y) = \begin{pmatrix} y \\ y \end{pmatrix} \end{pmatrix} \begin{pmatrix} y \\ y \end{pmatrix} \begin{pmatrix} y \\ y \end{pmatrix} \begin{pmatrix} y \\ y \end{pmatrix} \end{pmatrix} \begin{pmatrix} y \\ y \end{pmatrix} \begin{pmatrix} y \\ y \end{pmatrix} \begin{pmatrix} y \\ y \end{pmatrix} \end{pmatrix} \begin{pmatrix} y \\ y \end{pmatrix} \begin{pmatrix} y \\ y \end{pmatrix} \end{pmatrix} \begin{pmatrix} y \\ y \end{pmatrix} \begin{pmatrix} y \\ y \end{pmatrix} \begin{pmatrix} y \\ y \end{pmatrix} \end{pmatrix} \begin{pmatrix} y \\ y \end{pmatrix} \begin{pmatrix} y \\ y \end{pmatrix} \begin{pmatrix} y \\ y \end{pmatrix} \end{pmatrix} \begin{pmatrix} y \\ y \end{pmatrix} \begin{pmatrix} y \\ y \end{pmatrix} \begin{pmatrix} y \\ y \end{pmatrix}$$

Observe that: $F(0) = P(Y \le 0) = p(0) = 1/4$

$$F(0.3) = P(Y \le 0.3) = p(0) = 1/4$$
 (same), etc.

Also:
$$F(1) = P(Y \le 1) = p(0) + p(1) = 1/4 + 1/2 = 3/4$$

$$F(1.9) = P(Y \le 1.9) = p(0) + p(1) = 1/4 + 1/2 = 3/4$$
 (same)

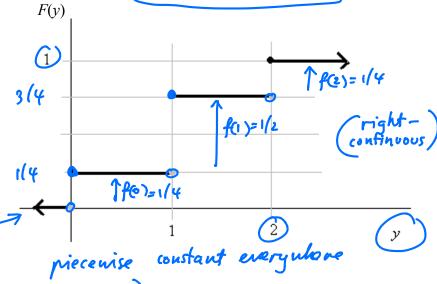
$$F(-3) = P(Y \le -3) = 0$$

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 $F(2.1) = P(Y \le 2.1) = 1$ etc.

Therefore Y's cdf is F(y) =

$$F(y) = \begin{cases} 0, & y < 0 \\ 1/4, & 0 \le y < 1 \\ 3/4, & 1 \le y < 2 \\ 1, & y \ge 2 \end{cases}$$



This is a step function, where each 'jump' corresponds to a probability. Eg, the jump at 2 is 1/4, which is the probability that Y equals 2. Note that F(0) equals 0.25, not 0.

Three properties of a cumulative distribution function

If F(v) is a cdf then:

- $F(y) \to 0$ as $y \to -\infty$ $F(y) \to 1$ as $y \to +\infty$ F(y) is nondecreasing.

F(y) is right continuous, meaning that $\lim_{\delta \downarrow 0} F(y+\delta) = F(y)$. Also, 4. (In Example 1 this corresponds to the fact that F(0) = 0.25, not 0.)

Definition of a continuous random variable

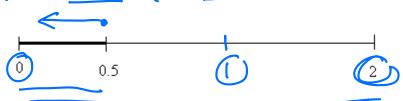
A random variable is said to be *continuous (cts)* if its cdf is continuous (no jumps (everywhere).

For instance, Y in Example 1 is *not* a continuous rv. (F(y)) is discontinuous at 0,1,2.)

Example 2 Let *Y* be a number chosen randomly between 0 and 2. Find Y's cdf. Is Y a cts rv?

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$$F(0.5) = P(Y \le 0.5) = 0.5/2 = 0.25$$

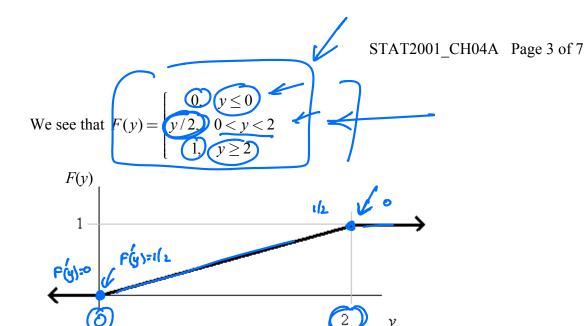


$$F(1) = P(Y \le 1) = 1/2$$

 $F(1.5) = P(Y \le 1.5) = 1.5/2 = 0.75$, etc.

Also:
$$F(-1) = P(Y \le -1) = 0$$

 $F(4) = P(Y \le 4) = 1$, etc.



Observe that F(y) is continuous everywhere (ie for all y between $-\infty$ and ∞). Hence Y is a continuous random variable.

The probability density function of a continuous random variable

What is the probability that Y equals 1? Answer: P(Y = 1) = 0.

(There is an uncountably infinite number of possible values of Y, and they are all equally likely; so each one occurs with probability 0.

Also, this follows from there being no 'jump' at y = 1 in Y's cdf, F(y).)

In fact,
$$P(Y=y) = 0$$
 for all y. $\in \mathbb{R}$

It follows that the earlier definition of a pdf (ie, p(y) = P(Y = y)) is now useless.

New definition: Suppose that Y is a continuous random variable with cdf F(y).

Then Y's probability density function (pdf) is

$$f(y) = F'(y)$$
 (= $dF(y)/dy$ = the derivative of $F(y)$).

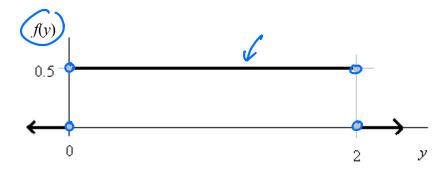
Example 3 Find Y's pdf in Example 2.

$$f(y) = \frac{dF(y)}{dy} = \begin{cases} \frac{d0}{dy} = 0, & y < 0 \\ \frac{d(y/2)}{dy} = \frac{1}{2}, & 0 < y < 2 \\ \frac{d0}{dy} = 0, & y < 0 \end{cases}$$

Note that f(y) is undefined at y = 0, 2.

(There are two different derivatives at each of these values.)

Eg, at y = 0, the left derivative is 0 and the right derivative is 1/2.



Observe that f(y) is the slope function of $\underline{F(y)}$. Eg, slope of F(y) at 0.1 is f(0.1) = 1/2.

Some simplifications

0.5

It is conventional to take undefined values as 0.

Also, we will not bother to always indicate exactly where a pdf is zero, nor where a cdf is 0 or 1. (These details have no effect on calculations when considering a purely *continuous* distribution. However, when dealing with a mixed distribution, as discussed in the optional Section 4.11, these details are important.)

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Graphs will also be simplified.

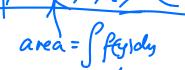
Thus we may write: F(y) = y/2, 0 < y < 2 f(y) = 1/2, 0 < y < 2.

Two properties of a continuous probability density function

Observe that no value of f(y) above is negative.

Also, the area under f(y) equals 1 (area = 2(1/2) = 1).

These properties hold for the pdf of every continuous random variable.



If f(y) is the pdf of a cts rv then:

- 1. $f(y) \ge 0$ for all y (NB: f(y) can be greater than 1.)
- 2. $\int f(y)dy = 1$. (NB: By default, the integral is over the whole

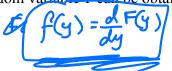
real line; thus it could also be written $\int f(y)dy$ or $\int f(y)dy$.

Observe in the last two figures that $\underline{F(0.5)} = 1/4$ is the same as the area under $\underline{f(y)}$ to the left of 0.5.

This fact may be written $F(0.5) = \int_{-\infty}^{0.5} f(y)dy$, or equivalently, $F(0.5) = \int_{-\infty}^{0.5} f(t)dt$.

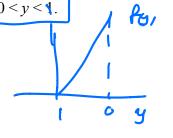
In general, the cdf F(y) of a continuous random variable Y can be obtained from its pdf f(y) via the equation

 $F(y) = \int_{0}^{\infty} f(t)dt.$



Example 3 Suppose that Y has pdf f(y) = 2y, 0 < y < 1. Find Y's cdf.

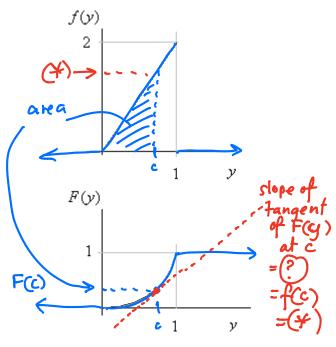
 $F(y) = \int_{0}^{y} 2t dt = \left[t^{2}\right]_{t=0}^{y} = y^{2} - 0^{2}$



So *Y*'s cdf is $F(y) = \sqrt{y^2}, 0 < y < 1$.

Note that we could now also 'switch back' to the pdf via differentiation:

$$f(y) = F'(y) = 2y, 0 < y < 1.$$



Consider any value c of Y ($c \in [0,1]$)

The area under f(y) to the left of c is $F(c) = c^{2}$

The slope of F(y) at y = c is F'(c) = f(c).

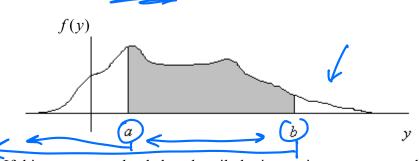
Eg: Area under f(y) to left of y = 1/2 is $F(1/2) = (1/2)^{2} = 1/4$ Slope of F(y) at y = 1/2 is $f(1/2) = 2 \times 1/2 = 1.$

The slopes of F(y) at 0 and 1 are 0 and 2.



Recall that P(Y = y) = 0 for all $y \in \mathbb{R}$ It follows that $P(Y \le y) = P(Y < y)$, $P(Y \ge y) = P(Y > y)$, etc.

The probability $P(a \le Y \le b)$ is the area under Y's pdf between a and b.



If this area cannot be deduced easily by inspection, we may need to do an integral:

$$P(a < Y < b) = \int_{a}^{b} f(y)dy.$$

However if we know Y's cdf, then we can instead use the formula:

$$P(a < Y < b) = F(b) - F(a)$$
.

(This is because P(a < Y < b) = P(Y < b) - P(Y < a), assuming that *Y* is a continuous random variable.)

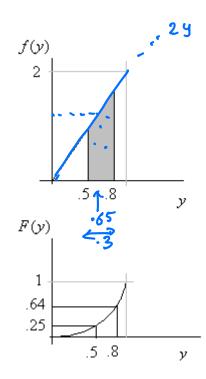
For example, in Eg 3, where f(y) = 2y, 0 < y < 1, what is P(0.5 < Y < 0.8)?

Solution 1:

This probability is the area of the shaded region below:

$$0.3 \times 2 \times 0.65 = 0.39$$

 $(0.65 \text{ is midway between } 0.5 \text{ and } 0.8, \text{ and } 2 \times 0.65 \text{ is the value of } f(y) \text{ at that point)}.$



$$P(0.5 < Y < 0.8) = \int_{0.5}^{0.8} f(y)dy = \int_{0.5}^{0.8} (2y)dy = \left[y^2 \Big|_{y=0.5}^{0.8} \right]$$
$$= 0.8^2 - 0.5^2 = 0.64 - 0.25 = 0.39$$

Solution 3:

$$P(0.5 < Y < 0.8) = F(0.8) - F(0.5) = 0.8^2 - 0.5^2 = 0.64 - 0.25 = 0.39$$
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