

two dice are rolled

Example 3.2.4 Return to the dice rolling experiment.

Let Y be the maximum that either die shows in a single roll:
 $Y(a, b) = \max(a, b)$.

For example, $Y(3, 5) = 5$.

$$Y(2, 6) = 6$$

$$Y(4, 4) = 4$$

(a) What is the range of Y ?

$$\{1, 2, 3, 4, 5, 6\}$$

(b) What is $P(Y = y)$ for each y in the range of Y ?

$$P(Y=1) = \frac{1}{36}$$

$$P(Y=2) = \frac{3}{36}$$

$$P(Y=3) = \frac{5}{36}$$

$$P(Y=4) = \frac{7}{36}$$

$$P(Y=5) = \frac{9}{36}$$

$$P(Y=6) = \frac{11}{36}$$

11	12	13	14	15	16
21	22	23	24	25	26
31	32	33	34	35	36
41	42	43	44	45	46
51	52	53	54	55	56
61	62	63	64	65	66

(c) Find a formula for the probability distribution of Y .

$$P(Y=y) = \frac{2y-1}{36}$$

Example 3.2.5 Check whether the function given by

$$f(x) = \frac{x+2}{25},$$

for $x = 1, 2, 3, 4, 5$ can serve as the probability distribution of a discrete random variable.

Check two things: 1) Are all $f(x)$ values non-negative? 2) Do the values of $f(x)$ add up to 1?

1) Yes, they are all non-negative.

$$2) f(1) + f(2) + f(3) + f(4) + f(5) = \frac{3}{25} + \frac{4}{25} + \frac{5}{25} + \frac{6}{25} + \frac{7}{25} = \frac{25}{25} = 1$$

So, yes the given function can serve as the probability distribution for a discrete random variable.

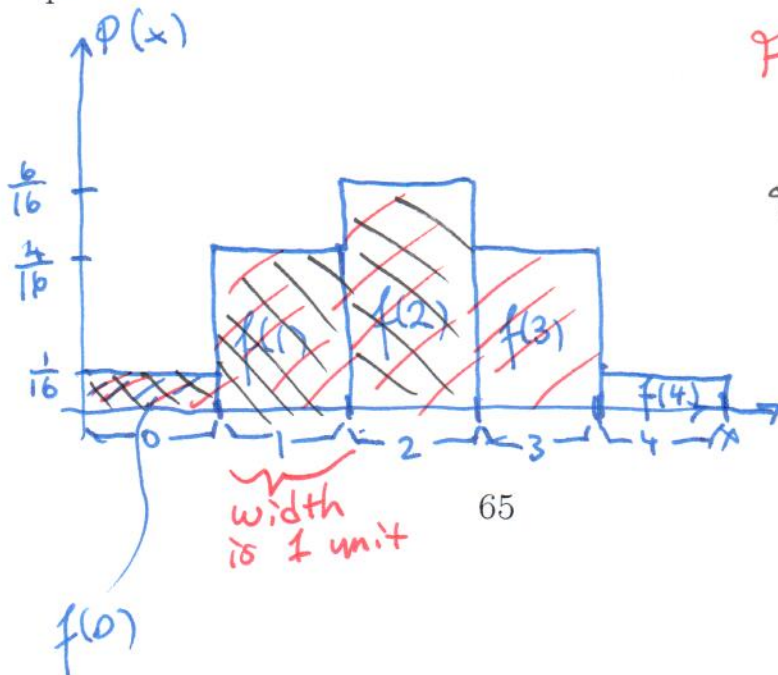
Probability distributions for a random variable, say X , may be represented graphically by means of a **probability histogram**.

Each rectangle corresponds to a value for X , its height is $P(X = x)$, and its width is 1, so that the area of each rectangle equals $P(X = x)$. The total area of the histogram is 1.

The probability histogram below is for the number of heads in 4 coin flips.

f:

$$\begin{aligned} P(X=0) &= \frac{1}{16} \\ P(X=1) &= \frac{4}{16} \\ P(X=2) &= \frac{6}{16} \\ P(X=3) &= \frac{4}{16} \\ P(X=4) &= \frac{1}{16} \end{aligned}$$



$$F(3) = f(0) + f(1) + f(2) + f(3)$$

$$F(2) = f(0) + f(1) + f(2)$$

$$F(3) - F(2) = f(3)$$

3.3 Cumulative Distribution

In many problems we are interested in the probability that the value of a random variable is less than or equal to (or “at most”) some real number x . i.e. $P(X \leq x)$.

If X is a discrete random variable with probability distribution f , the function given by

$$F(x) = P(X \leq x) = \sum_{t \leq x} f(t)$$

for $x \in (-\infty, \infty)$, is called the cumulative distribution of X (also called the **distribution function**).

Example 3.3.1 Let X be the random variable that counts the number of heads in 4 coin flips.

$$f(2) = \frac{6}{16} \text{ while}$$

$$F(2) = f(0) + f(1) + f(2) = \frac{1}{16} + \frac{4}{16} + \frac{6}{16} = \frac{11}{16}$$

$$f(0) = \frac{1}{16} \quad f(1) = \frac{4}{16} \quad f(2) = \frac{6}{16}$$

$$f(3) = \frac{4}{16} \quad f(4) = \frac{1}{16}$$

~~The corresponding columns of the probability histogram are as follows.~~

Back to a previous example:

Example 3.3.2 Two socks are selected at random and removed in succession from a drawer containing five brown socks and three green socks.

Let X be the random variable that counts the number of brown socks selected.

We found these values earlier on:

Element of sample space	Probability	x
BB	$\frac{20}{56}$	2
BG	$\frac{15}{56}$	1
GB	$\frac{15}{56}$	1
GG	$\frac{6}{56}$	0

The probability distribution f is given by

$$f(x) = \begin{cases} \frac{20}{56} & \text{for } x = 2 \\ \frac{30}{56} & \text{for } x = 1 \\ \frac{6}{56} & \text{for } x = 0 \end{cases}$$

$$\begin{aligned} F(0) &= f(0) = \frac{6}{56} \\ F(1) &= f(0) + f(1) = \frac{6}{56} + \frac{30}{56} = \frac{36}{56} \\ F(2) &= f(0) + f(1) + f(2) = \frac{6}{56} + \frac{30}{56} + \frac{20}{56} = 1 \end{aligned}$$

Find $F(0)$, $F(1)$, $F(2)$, and express the cumulative distribution (distribution function) $F(x)$ as a piece-wise defined function.

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$$F(x) = \begin{cases} 0 & x < 0 \\ \frac{6}{56} & 0 \leq x < 1 \\ \frac{36}{56} & 1 \leq x < 2 \\ \frac{56}{56} = 1 & 2 \leq x \end{cases}$$

Example 3.3.3 Suppose a random variable X has range $\{1, 2, 3, 4\}$. Define f by

$$f(1) = \frac{1}{4}, \quad f(2) = \frac{1}{2}, \quad f(3) = \frac{1}{8}, \quad f(4) = \frac{1}{8}$$

(a) Show that f is a valid probability distribution for X .

All values of f are non-negative.

$$f(1) + f(2) + f(3) + f(4) = \frac{1}{4} + \frac{1}{2} + \frac{1}{8} + \frac{1}{8} = \frac{8}{8} = 1$$

So, f is a valid prob. distribution.

(b) Find the cumulative distribution (distribution function) for X .

$$F(1) = f(1) = \frac{1}{4}$$

$$F(2) = f(1) + f(2) = \frac{1}{4} + \frac{1}{2} = \frac{3}{4}$$

$$F(3) = f(1) + f(2) + f(3) = \frac{1}{4} + \frac{1}{2} + \frac{1}{8} = \frac{7}{8}$$

$$F(4) = f(1) + f(2) + f(3) + f(4) = \frac{1}{4} + \frac{1}{2} + \frac{1}{8} + \frac{1}{8} = 1$$

$$F(x) = \begin{cases} 0 & x < 1 \\ \frac{1}{4} & 1 \leq x < 2 \\ \frac{3}{4} & 2 \leq x < 3 \\ \frac{7}{8} & 3 \leq x < 4 \\ 1 & 4 \leq x \end{cases}$$

Theorem 3.3.4 The cumulative distribution $F(x)$ satisfies

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

2. If $a < b$ then $F(a) \leq F(b)$ for any $a, b \in \mathbb{R}$.



Theorem 3.3.5 If the range of a random variable X consists of the values $x_1 < x_2 < \dots < x_n$, then $f(x_1) = F(x_1)$ and

$$\underline{f(x_i) = F(x_i) - F(x_{i-1})}$$

for $i = 2, 3, \dots, n$.

~~Let's see this on a probability histogram.~~

~~Let's see this on a probability histogram.~~

$$F(x_2) = f(x_1) + f(x_2)$$

$$F(x_3) = f(x_1) + f(x_2) + f(x_3)$$

$$F(x_3) - F(x_2) = f(x_3)$$

Example 3.3.6 The cumulative distribution for a discrete random variable X is given by

$$F(x) = \begin{cases} 0 & \text{for } x < -2 \\ \frac{4}{18} & \text{for } -2 \leq x < -1 \\ \frac{7}{18} & \text{for } -1 \leq x < 0 \\ \frac{12}{18} & \text{for } 0 \leq x < 1 \\ 1 & \text{for } x \geq 1 \end{cases}$$

$$(F(-3) = 0)$$

$$F(-2) = \frac{4}{18}$$

$$F(-1) = \frac{7}{18}$$

$$F(0) = \frac{12}{18}$$

$$F(1) = \frac{18}{18}$$

Find the probability distribution for X .

$$f(1) = F(1) - F(0) = \frac{18}{18} - \frac{12}{18} = \frac{6}{18}$$

$$f(0) = F(0) - F(-1) = \frac{12}{18} - \frac{7}{18} = \frac{5}{18}$$

$$f(-1) = F(-1) - F(-2) = \frac{7}{18} - \frac{4}{18} = \frac{3}{18}$$

$$f(-2) = F(-2) - F(-3) = \frac{4}{18} - 0 = \frac{4}{18}$$

$$f(x) = \begin{cases} \frac{4}{18} & x = -2 \\ \frac{3}{18} & x = -1 \\ \frac{5}{18} & x = 0 \\ \frac{6}{18} & x = 1 \end{cases}$$

3.4 Continuous Random Variables

3.4.1 Probability Density Function (p.d.f)

On a 100 km stretch of rural road we are concerned with the possibility that a deer might cross.

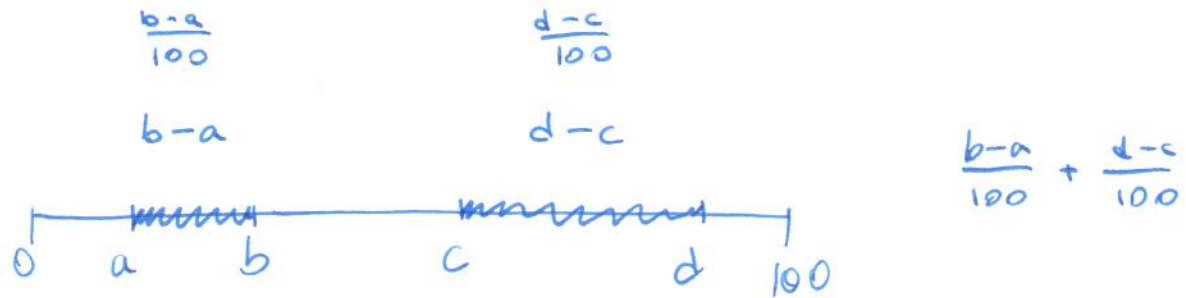
We are interested in the probability that it will occur at a given location or stretch of the road. The sample space for this experiment consists of all points in the interval from 0-100.

Suppose the probability that a deer crosses in any stretch of road is the length of that section divided by 100.

So, from point a to point b with $0 \leq a, b \leq 100$, is the interval $[a, b]$ and its length is given by $b - a$. So, its probability is

$$P([a, b]) = \frac{b - a}{100}.$$

Handwritten notes:
→ length of the stretch from a to b (pointing to $b - a$)
→ total length (pointing to 100)



The probability of any two or more non-overlapping intervals can be found by summing the probabilities of the connected components.

Thus the probability measure proposed here has non-negative values, assigns the entire sample space a probability of 1, and is countably additive; hence it satisfies our postulates of probability.

$$\frac{100 - 0}{100} = 1$$

We have taken the sample space to be any point on this stretch of road, and the random variable X here is the function that assigns that point to a real number in the interval $[0, 100]$. This is an example of a **continuous random variable**.

We can give the probability that X lies within an interval by

$$P(a \leq x \leq b) = \frac{b - a}{100} \quad [a, b]$$

for $a < b$, however the probability that X is any single point is zero.

In the case of a continuous random variable, probabilities cannot simply be assigned to every outcome as is done with a discrete random variable.

Therefore a **continuous random variable** must be accompanied by a **probability density function** in order to compute probabilities.

A positive-valued function f defined on \mathbb{R} is called a **probability density function** for continuous random variable X , if and only if

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

for any $a, b \in \mathbb{R}$ with $a \leq b$. These are also called "p.d.f's" for short.

In the deer crossing example, the p.d.f. for X is $f(x) = \frac{1}{100}$.

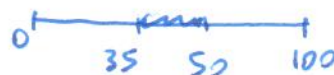
For example

$$\begin{aligned} P(35 \leq X \leq 50) &= \int_{35}^{50} \frac{1}{100} dx \\ &= \left. \frac{x}{100} \right|_{35}^{50} = \frac{50 - 35}{100} = \frac{15}{100}. \end{aligned}$$

Notice that $f(r)$ does not give the probability that $X = r$.

$$\begin{aligned} &\int_{35}^{50} f(x) dx \\ &= \int_{35}^{50} \frac{1}{100} dx \\ &= \left. \frac{x}{100} \right|_{35}^{50} \\ &= \frac{50}{100} - \frac{35}{100} \\ &= \frac{15}{100} \end{aligned}$$

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$$50 - 35 = 15$$

Prob. that
a deer might cross between
35 and 50 is $\frac{15}{100}$

Let X be a continuous random variable. By properties of integrals it follows that

Theorem 3.4.1 *If $a, b \in \mathbb{R}$ with $a \leq b$ then*

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

$$[a, b] \quad [a, b) \quad (a, b] \quad (a, b)$$

(informally: because the contribution of a single point to an integral is 0)

From the postulates of probability we obtain the following result:

Theorem 3.4.2 *A function f can serve as a probability density function for X only if it satisfies*

1. $f(x) \geq 0$ for all $x \in \mathbb{R}$.

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

(very similar to the discrete case)

continuous random variable

Example 3.4.3 If X has probability density function

$$f(x) = \begin{cases} k \cdot e^{-3x} & \text{for } x > 0 \\ 0 & \text{elsewhere} \end{cases}$$

find k and $P(0.5 \leq X \leq 1)$.

First we need to find what k is.

Since f is given to be a probability density function, it satisfies Condition 2 of the previous theorem.

Solve for k using Condition 2. from the theorem.

$$\begin{aligned} 1 &= \int_{-\infty}^{\infty} f(x) \, dx = \int_{-\infty}^0 0 \, dx + \int_0^{\infty} k \cdot e^{-3x} \, dx \\ &= \lim_{c \rightarrow \infty} k \left. \frac{e^{-3x}}{(-3)} \right|_0^c \\ &= \lim_{c \rightarrow \infty} k \frac{e^{-3c}}{(-3)} - k \frac{e^{-3(0)}}{(-3)} \\ &= \frac{k}{3}. \quad (\text{since } \lim_{r \rightarrow \infty} e^{-r} = 0) \end{aligned}$$

Thus $k = 3$. Now we can compute

$$\begin{aligned} P(0.5 \leq X \leq 1) &= \int_{0.5}^1 f(x) \, dx = \int_{0.5}^1 3e^{-3x} \, dx = -e^{-3x} \Big|_{0.5}^1 \\ &= -e^{-3} - (-e^{-1.5}) \approx 0.1733 \end{aligned}$$

From Theorem 3.4.2 on page 74 (part 2.) we know that

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

$$f(x) = \begin{cases} k \cdot e^{-3x} & x > 0 \\ 0 & x \leq 0 \end{cases} \quad \text{p.d.f.}$$

$$\Rightarrow \int_{-\infty}^0 0 dx + \int_0^{\infty} k \cdot e^{-3x} dx = 1$$

improper
integrals

$$\Rightarrow \int_0^{\infty} k \cdot e^{-3x} dx = 1 \Rightarrow \lim_{c \rightarrow \infty} \int_0^c k \cdot e^{-3x} dx = 1$$

$$\Rightarrow \lim_{c \rightarrow \infty} \left(\frac{k}{-3} e^{-3x} \Big|_0^c \right) = 1$$

an antiderivative
for $k \cdot e^{-3x}$ is $\frac{k}{-3} e^{-3x}$

$$\Rightarrow \lim_{c \rightarrow \infty} \left(\frac{k}{-3} e^{-3c} - \frac{k}{-3} e^{-3 \cdot 0} \right) = 1$$

$$\Rightarrow -\frac{k}{-3} e^0 = 1 \Rightarrow \frac{k}{3} \cdot 1 = 1 \Rightarrow \underline{k=3}$$

Next, we find $P(0.5 \leq X \leq 1)$.

That is, we calculate $\int_{0.5}^1 f(x) dx = \int_{0.5}^1 3 \cdot e^{-3x} dx$

an antiderivative
for $3 \cdot e^{-3x}$ is

$$= -e^{-3x} \Big|_{0.5}^1 = -e^{-3 \cdot 1} - (-e^{-1.5})$$

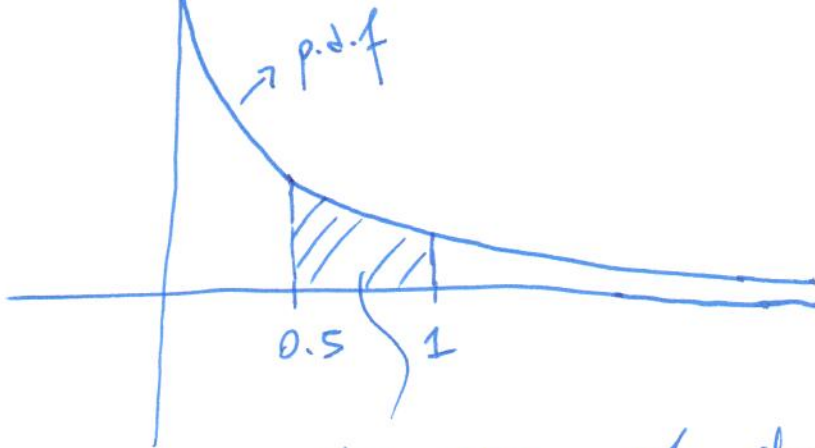
$$3 \cdot \frac{e^{-3x}}{-3} = -e^{-3x}$$

$$= -e^{-3} + e^{-1.5} \approx 0.1733$$

probability
density function

Graph of $3e^{-3x}$ is given below.

The shaded area denotes $P(0.5 \leq X \leq 1)$.



the area of the
shaded region is $= P(0.5 \leq x \leq 1)$

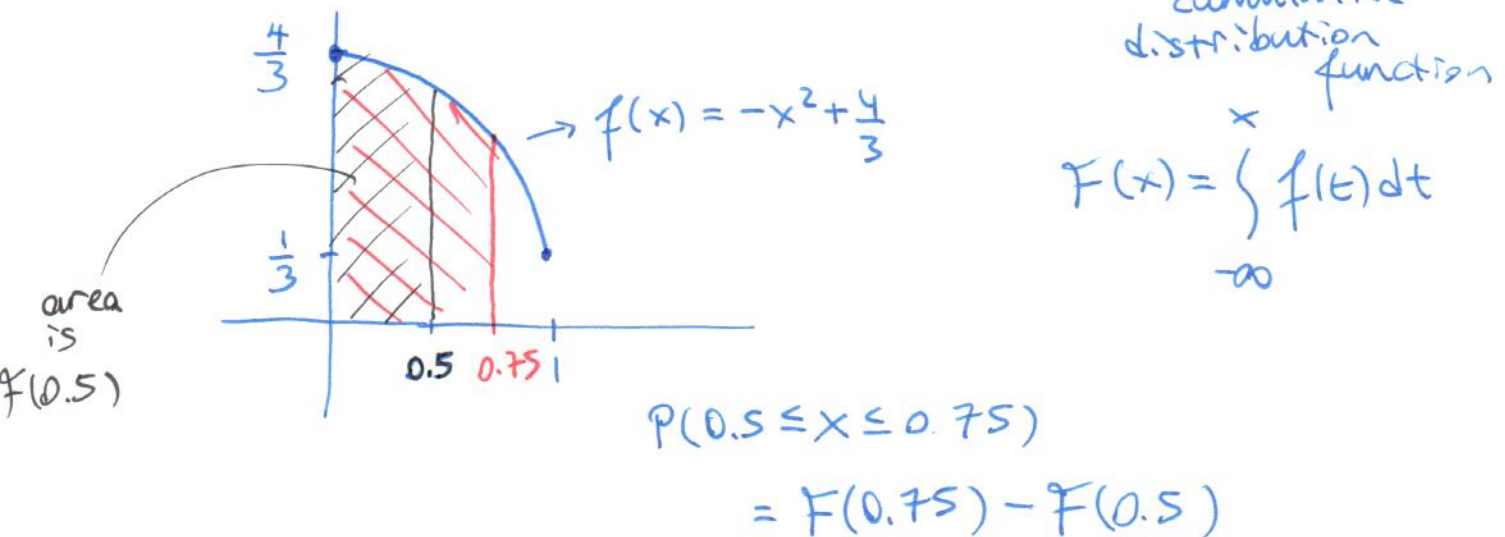
3.4.2 Cumulative Distribution Function of a Continuous Random Variable

Let X ^{be} ~~is~~ a continuous random variable with probability density function f . Then the function

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

for all $x \in \mathbb{R}$, is called the **cumulative distribution function** of X .

Example 3.4.4 Random variable X with p.d.f. $f(x) = -x^2 + \frac{4}{3}$ for $0 \leq x \leq 1$ and 0 elsewhere. (p.d.f. plotted in ~~red~~ ^{blue})



Cumulative distribution function is $F(x) = \int_{-\infty}^x f(t) dt$.

Shade the areas representing the values $F(0.5)$ and $F(0.75)$ respectively.

From the properties of integrals we have the following.

Theorem 3.4.5 *If continuous random variable X has probability density function $f(x)$ and cumulative distribution function $F(x)$ then*

$$P(a \leq X \leq b) = F(b) - F(a)$$

for any $a, b \in \mathbb{R}$ with $a \leq b$, and

$$f(x) = \frac{d}{dx}F(x)$$

where the derivative exists.

Using the previous example with p.d.f. $f(x) = -x^2 + \frac{4}{3}$ for $0 \leq x \leq 1$, and 0 elsewhere, we have:

$$P(0.25 \leq X \leq 0.75) = F(0.75) - F(0.25)$$

Let's see this considering the relevant shaded areas on the corresponding graphs.

done on the previous page

p.d.f

$$f(x) = \begin{cases} -x^2 + \frac{4}{3} & 0 \leq x \leq 1 \\ 0 & \text{elsewhere} \end{cases}$$

The cumulative distribution function is

$$F(x) = \int_{-\infty}^x f(t) dt = \int_0^x -t^2 + \frac{4}{3} dt = \dots \left(\frac{-t^3}{3} + \frac{4}{3}t \right) \Big|_{t=0}^{t=x}$$

because
f is 0
on $(-\infty, 0)$

$$= \frac{-x^3}{3} + \frac{4}{3}x - 0 = -\frac{x^3}{3} + \frac{4}{3}x$$

and its derivative is the probability density function

$$\frac{d}{dx}F(x) = \frac{d}{dx} \left(-\frac{x^3}{3} + \frac{4x}{3} \right) = \dots \frac{3x^2}{3} + \frac{4}{3} = -x^2 + \frac{4}{3}$$