

# RV: CUMULATIVE DISTRIBUTION FUNCTION

Dr. Aleksandra Kaszubowska-Anandarajah

[anandara@tcd.ie](mailto:anandara@tcd.ie)

Some material has been adopted from  
the Pennsylvania State University online course STAT 414

# Previous lecture

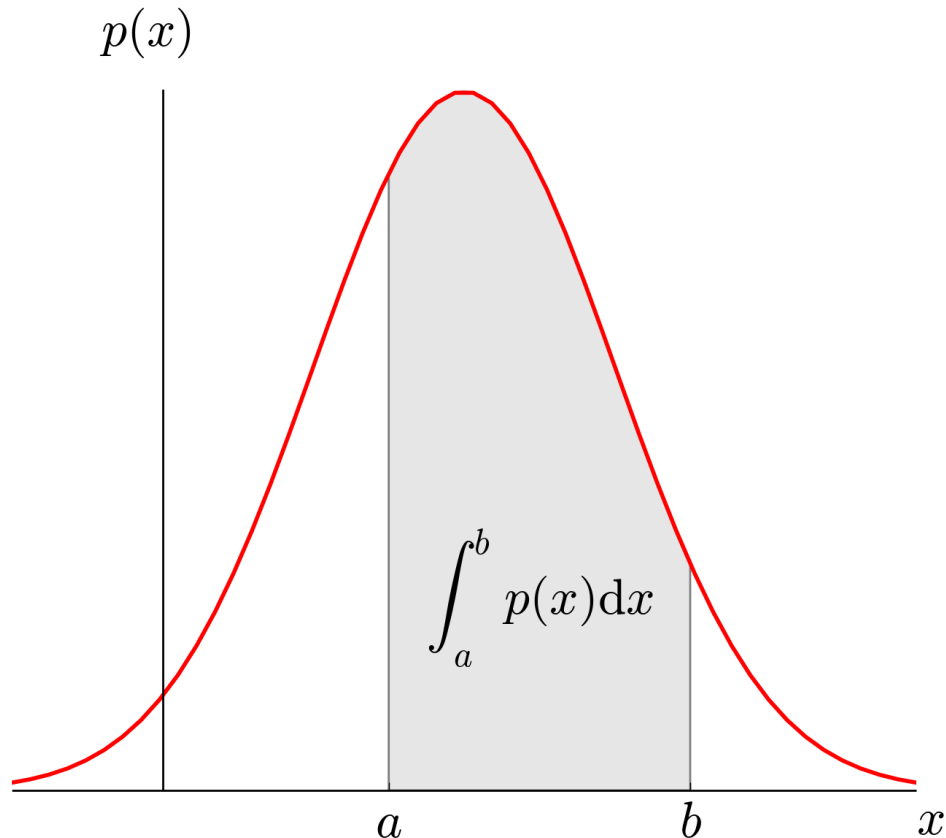
- Law of total probability
- Bayes theorem
- Statistical independence
- Random variable
- Probability mass function

# Today's lecture

- Probability density function
- Cumulative distribution function
- Expected value and its properties

# Probability Density Function (PDF)

- For continuous RV the probability that  $X=x$  is zero.
- Instead we seek the probability that  $X$  falls within an interval  $a < X < b$



- The probability that the continuous RV  $X$ , is between  $a$  and  $b$  corresponds to the area under the curve representing the probability density function between the limits  $a$  and  $b$ .
- For discrete RV it was a sum off  $\sum p(X=x_i)$

## Properties of PDF

PDF is an integrable function with the following properties:

1.  $f(x) > 0$ , for all  $x$  in  $S$
2. The area under the curve  $f(x)$  in the support  $S$  is 1, that is:  $\int_S f(x)dx = 1$
3. If  $f(x)$  is the **PDF** of  $x$ , then the probability that  $x$  belongs to some interval, is given by the integral of  $f(x)$  over that interval, that is:

$$P(x \in A) = \int_A f(x)dx$$

# PDF - Example

Let  $f(x) = 3x^2$  for  $0 < x < 1$

1. Verify that  $f(x)$  is a valid PDF
2. Calculate  $P(1/2 < X < 1)$  &  $P(X=1/2)$ .

Note:  $f(x)$  is not a probability  $\Rightarrow$  e.g.: for  $x=0.7 \Rightarrow f(x)=1.47 \Rightarrow f(x)>1!$

Solution:

1.

a) Is  $f(x)$  nonnegative for all  $x \in S$  i.e.  $0 < x < 1$ ?

Answer:  $3x^2$  is always positive

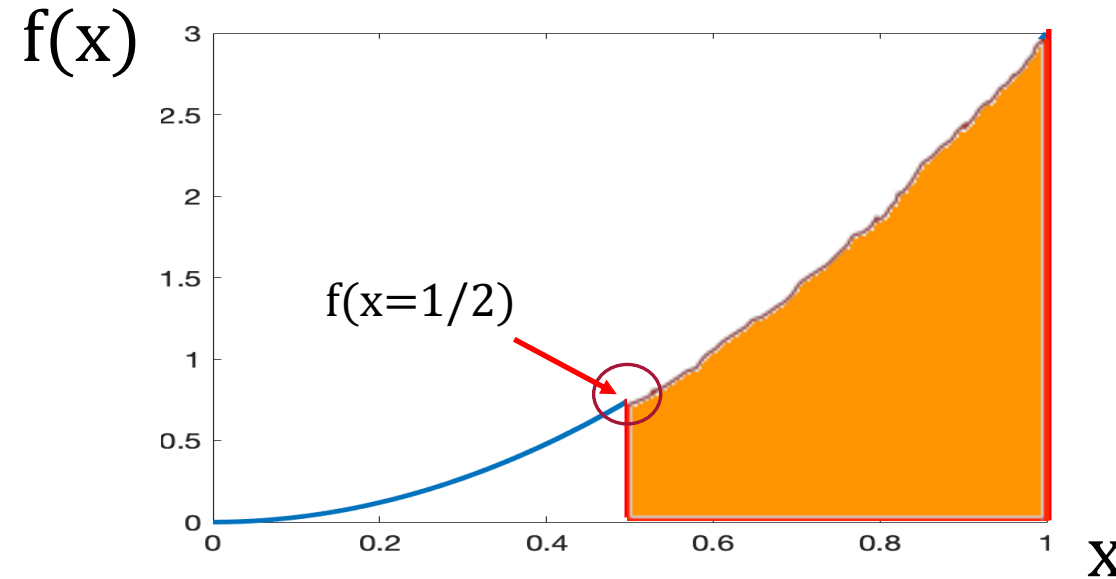
b)  $\int_S f(x) dx = 1 \Rightarrow \int_0^1 3x^2 dx = [x^3]_0^1 = 1 - 0 = 1$

Answer:  $f(x)$  integrated over the entire support equals 1 i.e.  $f(x)$  is a valid PDF

# PDF - example 1

2.

a)  $P\left(\frac{1}{2} < x < 1\right)$



$$P\left(\frac{1}{2} < x < 1\right) = \int_{\frac{1}{2}}^1 3x^2 dx = [x^3]_{\frac{1}{2}}^1 = 1 - \left(\frac{1}{2}\right)^3 = 1 - \frac{1}{8} = \frac{7}{8}$$

b)  $P\left(x = \frac{1}{2}\right) = \int_{\frac{1}{2}}^{\frac{1}{2}} 3x^2 dx = [x^3]_{\frac{1}{2}}^{\frac{1}{2}} = \frac{1}{8} - \frac{1}{8} = 0$



# Grouped Frequencies

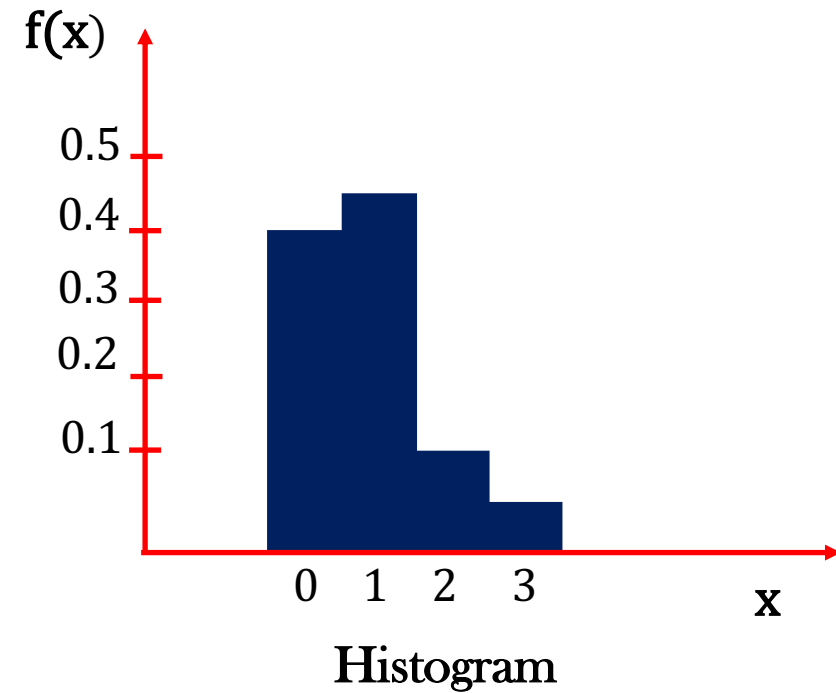
- For continuous RV, within a range, there is an infinite number of possible values
- Thus, representing the continuous data is easier by dividing the variate into intervals or classes and counting the frequency of occurrence for each class
- This is called the grouped frequency approach
- The width of the class is usually constant
- # of classes can be determined using empirical Sturges' rule:

$$\text{\#class intervals} \approx 1 + 3.3 \log_{10} N$$

N - total number of observations in the sample or population

# Grouped Frequencies

- The class boundaries must be clear with no gaps and no overlaps
- The relative frequency: ratio of the class frequency and the total of all the class frequencies
- Often represented as a histogram



# Cumulative Distribution Function

## CDF

$$F_X(s) = P(X \leq s)$$

- $F_X(s)$  is a nondecreasing function of  $s$ , for  $-\infty < s < \infty$
- The CDF ranges from 0 to 1 ( $F_X(s)$  is a probability!)
- Also called the distribution function.
- All probabilities concerning  $X$  can be stated in terms of  $F_X(s)$ .

# CDF of a discrete RV

## CDF of a discrete RV

$$F_X(s) = P(X \leq s) = \sum_{x \leq s} f(x)$$

- If  $X$  is a discrete random variable whose minimum value is  $z$ , then:

$$F_X(z) = P(X \leq z) = P(X=z) = f_X(z).$$

If  $c$  is less than  $z$ , then  $F_X(c) = 0$

- If the maximum value of  $X$  is  $b$ , then  $F_X(b) = 1$
- CDF of a discrete RV is a non-decreasing step function

# CDF - Example 1

PMF of a discrete random variable  $X$  is equal to:

$$f(x) = \frac{5-x}{10} \quad \text{for } x=1,2,3,4.$$

Find the CDF of  $X$ .

x	1	2	3	4
f(x)=P(X=x)				

# CDF - solution

The CDF is:  $F_X(s) = P(X \leq s)$

➤ For  $s=1$ ,  $P(X \leq 1) = P(X=1) = f(1) = \frac{5-s}{10} = \frac{4}{10}$

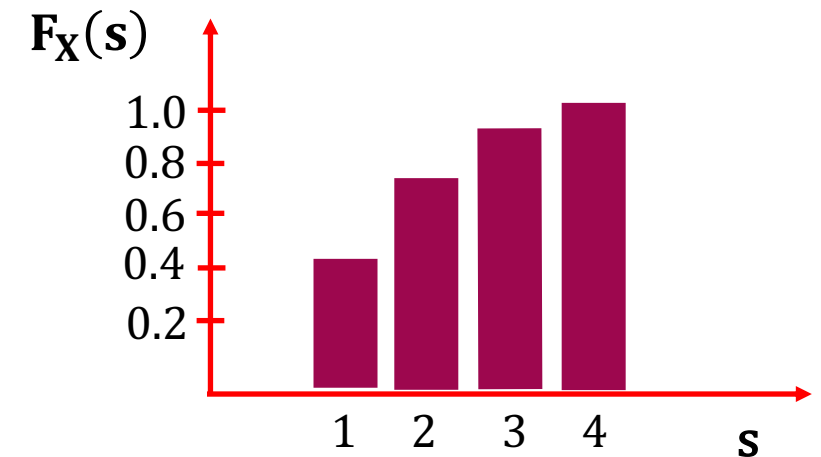
➤ For  $s=2$ ,  $P(X \leq 2) = P(X=1 \text{ or } X=2) = P(X=1) + P(X=2) = \frac{5-1}{10} + \frac{5-2}{10} = \frac{7}{10}$

➤ For  $s=3$ ,  $P(X \leq 3) = \frac{5-1}{10} + \frac{5-2}{10} + \frac{5-3}{10} = \frac{9}{10}$

➤ For  $s=4$ ,  $P(X \leq 4) = \frac{5-1}{10} + \frac{5-2}{10} + \frac{5-3}{10} + \frac{5-4}{10} = 1$

x	1	2	3	4
f(x)=P(X=x)	0.4	0.3	0.2	0.1

s	1	2	3	4
$F_X(s) = P(X \leq s)$	0.4	0.7	0.9	1



# CDF of a continues RV

## CDF of a continues RV

$$F_X(s) = P(X \leq s) = \int_{-\infty}^s f(x)dx$$

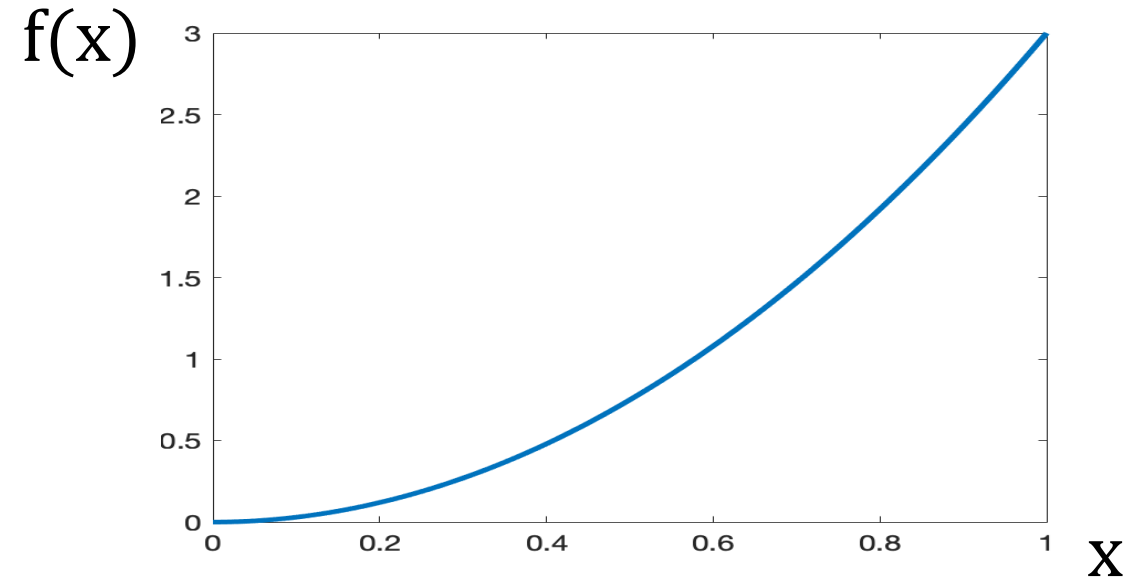
- CDF of a continues RV is a non-decreasing continues function

## CDF - Example 2

For the RV  $X$  with the probability density function:

$$f(x) = 3x^2 \text{ for } 0 < x < 1.$$

calculate the CDF





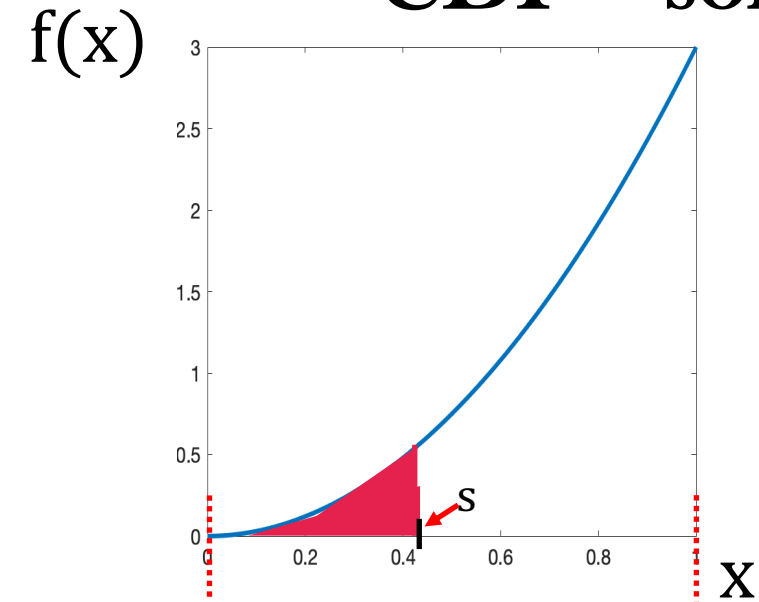
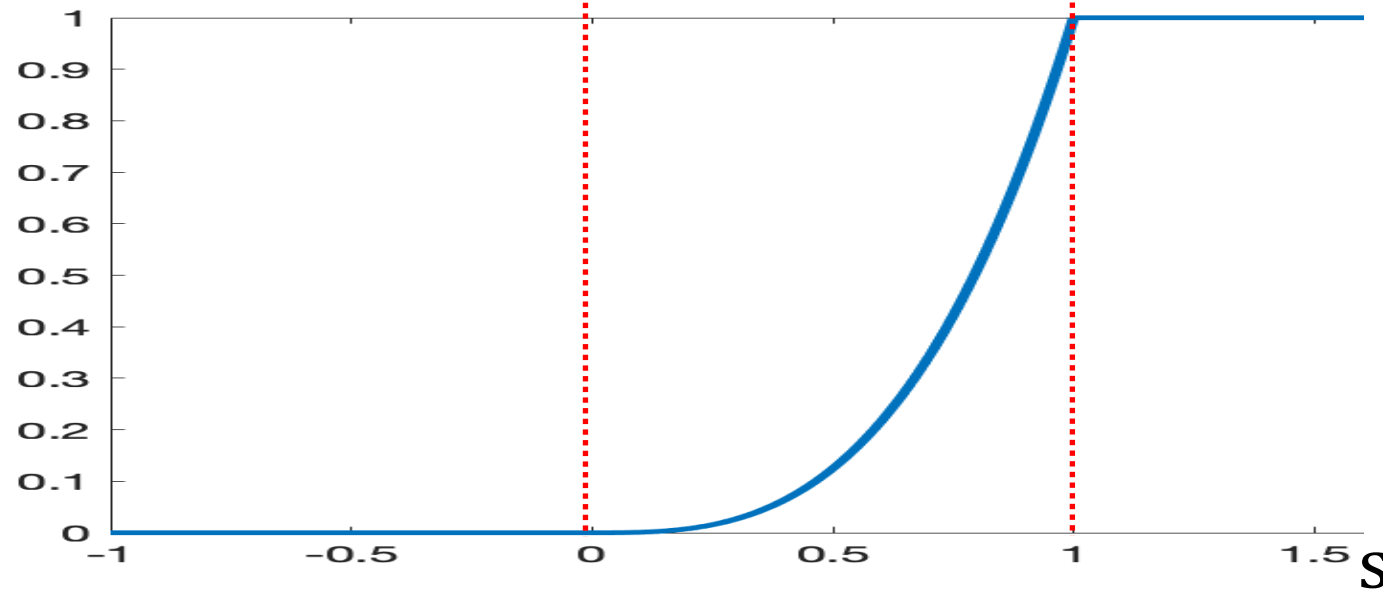
## CDF - solution

The CDF is:  $F_X(s) = P(X \leq s) = \int_{-\infty}^s f(x)dx$

$$F_X(s) = \int_0^s 3x^2 dx = [x^3]_0^s = s^3 \quad \text{for } 0 < s < 1$$

$$F_X(s) = \begin{cases} 0, & \text{for } s \leq 0 \\ s^3, & \text{for } 0 < s < 1 \\ 1, & \text{for } s \geq 1 \end{cases}$$

$F_X(s)$



# Example

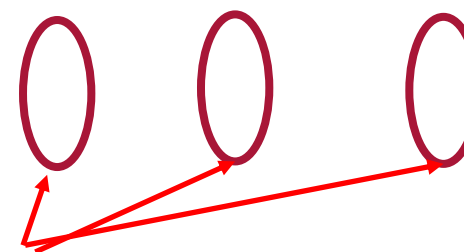
- Calculate average of the result of 36 dice throws

①	3	2	4	5	6	5	4	4	①	2	3
2	①	3	6	4	5	5	4	3	①	6	6
2	5	3	①	2	4	6	2	5	6	3	①

.....

$$\text{Avg}(x) = \frac{(1 + 1 + 1..+1) + (2 + 2..+2) + \dots (6 + 6 + \dots + 6)}{36} =$$

$$= \frac{\cancel{6}(1) + \cancel{6}(2) + \dots \cancel{6}(6)}{\cancel{36}}$$



Probabilities of throwing a given no on the dice for large no of observations

Remember that  $P(A) = \lim_{n \rightarrow \infty} \left( \frac{N_n(A)}{n} \right)$

# Expected Value of a RV

- Expected value - the mean of all possible results for an infinite number of trials
- Expected value is an average of the values weighted by their probabilities

## Expected value or a MEAN of a RV

$$\mu_X = E(X) = \sum_{x \in S} x f(x) \quad \text{for discrete RV}$$

$$\mu_X = E(X) = \int_{-\infty}^{\infty} x f(x) dx \quad \text{for continuous RV}$$

- $E(X)$  exists only if sum exists/integral converges absolutely ( $\neq \infty$ )

# Expected Value - Example 1

The probability that a 30-year-old man will survive a fixed length of time is 0.995. The probability that he will die during this time is thus  $1 - 0.995 = 0.005$ . An insurance company will sell him a \$20,000 life insurance policy for this length of time for a premium of \$200.00. What is the expected gain for the insurance company?

Answer:

If the man lives through the fixed length of time, the company's gain will be \$200.00. The probability of this is 0.995.

If the man dies during this time, the company's gain will be  $+\$200.00 - \$20,000.00 = -\$19,800.00$ . The probability of this is 0.005.

Thus the expected gain for the company is

$$E(X) = (\$200.00)(0.995) + (-\$19,800.00)(0.005) = \$199.00 - \$99.00 = \$100.00$$

# Expected Value of a Function of RV

## Expected value of a function of a RV

$$E(g(X)) = \sum_{x \in S} g(x)f(x) \quad \text{for discrete RV}$$

$$E(g(X)) = \int_{-\infty}^{\infty} g(x)f(x)dx \quad \text{for continuous RV}$$

## Expected Value - Example 2

Calculate  $E(g(X))$  for a function  $g(x) = \cos(x)$ ,

where  $X$  is a RV with following uniform PDF:

$$f(x) = \begin{cases} \frac{1}{2\pi} & \text{for } -\pi < x < \pi \\ 0 & \text{otherwise} \end{cases}$$

Solution:

$$E(X) = \int_{-\pi}^{\pi} \cos(x) \frac{1}{2\pi} dx$$

# Properties of Expected Value

## Properties of $E(X)$

1.  $E(c) = c$  where  $c$  is a constant
2.  $E(cg(X)) = cE(g(X))$
3.  $E(c_1g_1(X) + c_2g_2(X)) = c_1E(g_1(X)) + c_2E(g_2(X))$

Property 3 can be extended to:

$$E\left(\sum_i c_i g_i(X)\right) = \sum_i c_i E(g_i(X))$$

# Properties of Expected Value

Proof 1:

$$E(c) = \sum_{x \in S} cf(x) = c \sum_{x \in S} f(x) = c$$

=1 from the property 2 of PMF

Proof 2:

$$E(cg(X)) = \sum_{x \in S} cg(x)f(x) = c \sum_{x \in S} g(x)f(x) = cE(X)$$

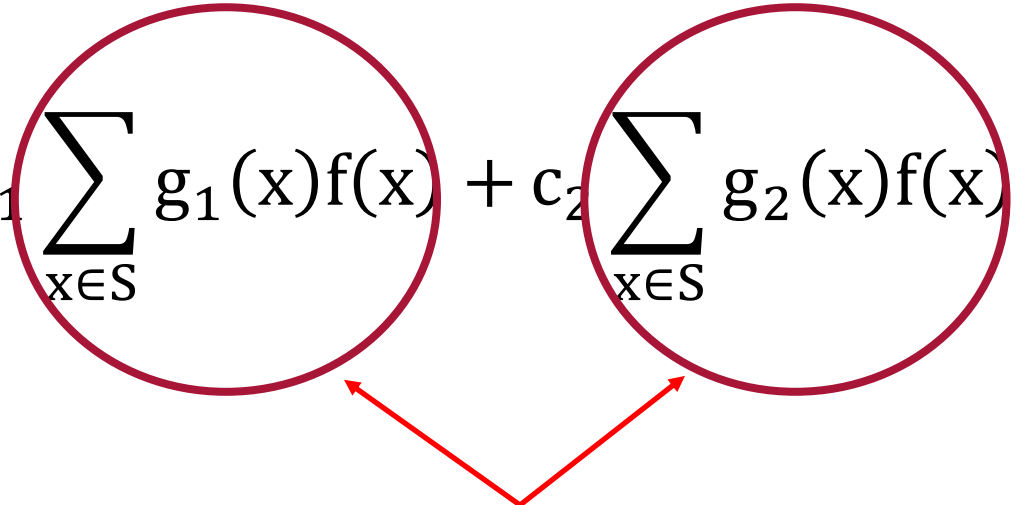
=E(X) from the definition



# Properties of Expected value

Proof 3:

$$E(c_1g_1(X) + c_2g_2(X)) = \sum_{x \in S} (c_1g_1(x) + c_2g_2(x))f(x) =$$

$$= \sum_{x \in S} c_1g_1(x)f(x) + \sum_{x \in S} c_2g_2(x)f(x) = c_1 \sum_{x \in S} g_1(x)f(x) + c_2 \sum_{x \in S} g_2(x)f(x)$$


$$= c_1E(g_1(X)) + c_2E(g_2(X))$$

$E(X)$  from the definition

# What have we learnt?

- Probability density function
- Cumulative distribution function
- Expected value and its properties