Random variables

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Definition

- A random variable, denoted by X, is a function that maps each outcome in the sample space of a random experiment to a real number.
- This mapping allows us to quantify outcomes in a way that facilitates mathematical analysis and probability calculations
- There are two types of random variables
 - Discrete random variable: countable and the values can be listed x_1, x_2, x_3 ...
 - Continuous random variable: not countable and cannot be listed, presented as interval

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Example

- In the experiment of flipping a coin two times, and X is a random variable represents the number of heads
 - what is the possible values of X

Solution

$$S = \{HH, HT, TH, TT\}$$

2 1 1 0

- x = 0, 1, 2
- we can ask what is the probability of x P(X = x), such that x is an outcome

$$P(X = 0) = \frac{1}{4}, P(X = 1) = \frac{2}{4}$$



Other examples

- X is a random number representing the number of customers who enter a store each hour.
 - X is considered a random variable because this number can vary and is subject to change
- Y is a random variable representing the service time in minutes of a customer by a specific server
- Z is a random variable representing the money you hold in your wallet in JOD
- W is a discrete random variable representing the number of cars expected to enter a gas station on a particular day of the month.

Probability mass function

- For a discrete random variable X with possible values $x_1, x_2, x_3...$ The probability mass function (PMF) is a function such that:
 - 1. $f(x_i) \ge 0$
 - 2. $\sum_{i=1}^{n} f(x_i) = 1$
- $f(x_i)$ is equal to $P(X = x_i)$,

Think of the probability of the color value in an image

Color V	0	1	2	3	•••	255
Probability	0.01	0.09	0.02	0.05		0.001

The some of these probabilities must equal 1, if not then it is not PMF

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Example

• given the following mass function:

x	-2	-1	0	1	2
f(x) = P(X = x)	1/8	2/8	2/8	2/8	1/8

• Find

1.
$$P(X \le 2)$$

2.
$$P(X \ge 0)$$

3.
$$P(-2 < X < 2)$$

4.
$$P(X < 0 \text{ or } X = 1)$$

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Cumulative Distribution Function (CDF)

• For a random variable X, the Cumulative Distribution Function, F(x), is defined for every number x as:

$$F(x) = P(X \le x)$$

• This means F(x) represents the probability that the random variable X will have a value less than or equal to x

$$F(x) = P(X \le x) = \sum_{x_i \le x} f(x_i)$$

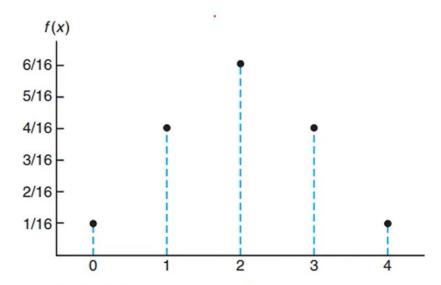
• When X is discrete

x	-2	-1	0	1	2
f(x) = P(X = x)	1/8	2/8	2/8	2/8	1/8
$F(x) = P(X \le x)$	1/8	3/8	5/8	7/8	8/8

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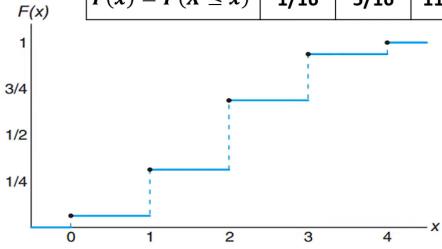
PMF vs CDF plot

x	0	1	2	3	4
f(x) = P(X = x)	1/16	4/16	6/16	4/16	1/16



Probability mass function plot.

x	0	1	2	3	4
f(x) = P(X = x)	1/16	4/16	6/16	4/16	1/16
$F(x) = P(X \le x)$	1/16	5/16	11/16	15/16	16/16



Discrete cumulative distribution function.

Illustration

• let us assume that x represent the number of typos in a given page of a book

x	0	1	2	3	4
f(x) = P(X = x)	1/16	4/16	6/16	4/16	1/16
$F(x) = P(X \le x)$	1/16	5/16	11/16	15/16	16/16

• Find:

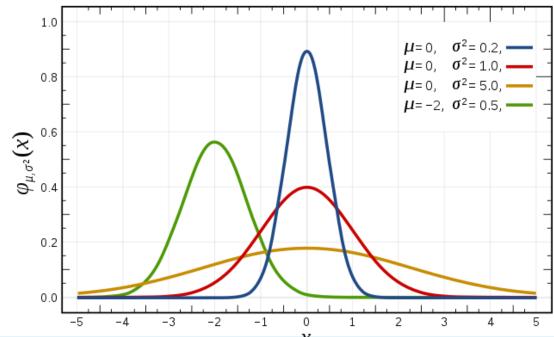
- 1. The probability of having less than or equal 2 errors
 - this can be found using the CDF F(2)=11/16
- 2. Errors fewer than or equal to 4 errors
 - *F*(4)=16/16
- 3. Errors in-between 1 and 3
 - f(1)+f(2)+f(3)=4/16+6/16+4/16=14/16
 - OR F(3)-F(0)=15/16-1/16=14/16

- Such data can be collected from real life.
- we will come later in this course on how to collect data and make such an analysis



Mean and Variance

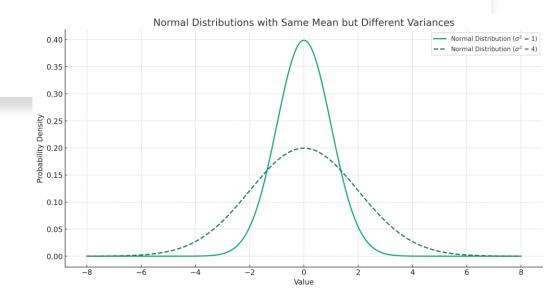
- Mean and variance are very important metrics used to summarize a probability distribution for a random variable.
- Mean represents the center of the probability distribution, while variance measures the dispersion of the distribution

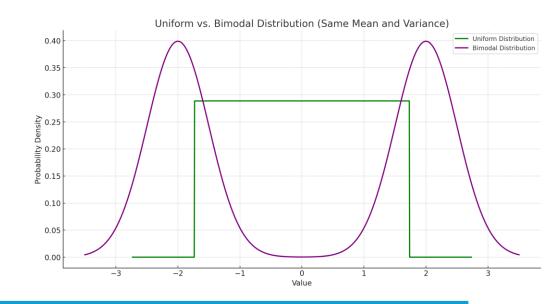




• Sometimes, the probability distribution has the same mean but different variance

 Or the same mean and same variance with different distribution





Mean/Variance/Standard deviation

• The mean μ (aka expected value E(X)) can be calculated using the following formula

$$\mu = E(X) = \sum_{x} x f(x)$$

• Variance σ^2 or V(X)

$$\sigma^{2} = E(X^{2}) - (E(X))^{2} = \sum_{x} x^{2} f(x) - \mu^{2}$$

Standard deviation

$$\sigma = \sqrt{\sigma^2}$$

Example

- Calculate the expected value and variance of the following distribution
- Expected value

$$E(X) = 1\left(\frac{1}{6}\right) + 2\left(\frac{1}{6}\right) + 3\left(\frac{1}{6}\right) + 4\left(\frac{1}{6}\right) + 5\left(\frac{1}{6}\right) + 6\left(\frac{1}{6}\right) = 3.5$$

Variance

$$E(X^2) = 1^2 \left(\frac{1}{6}\right) + 2^2 \left(\frac{1}{6}\right) + 3^2 \left(\frac{1}{6}\right) + 4^2 \left(\frac{1}{6}\right) + 5^2 \left(\frac{1}{6}\right) + 6^2 \left(\frac{1}{6}\right) = 15.16$$

$$V(X) = 15.16 - 3.5^2 = 2.92$$

Continuous Random variable

• If the range of the random variable *X* is an interval e.g. 1-2, then this random variable is continuous, as the numbers in this range are not countable and can not be listed

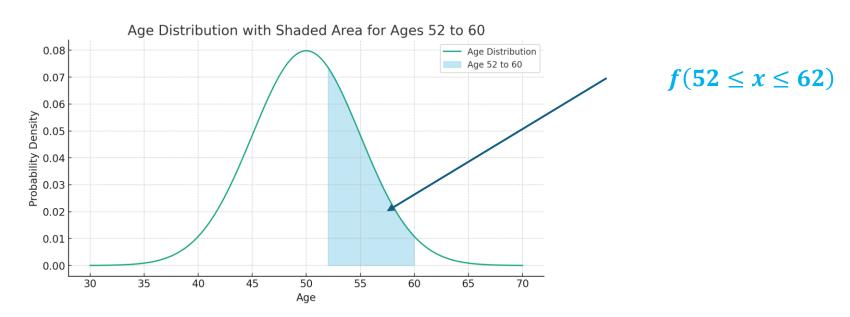
- In continuous we cannot find the probability of a single value due to the use of the integration and area-under-curve calculation
 - There is no area of a single line, its width is 0

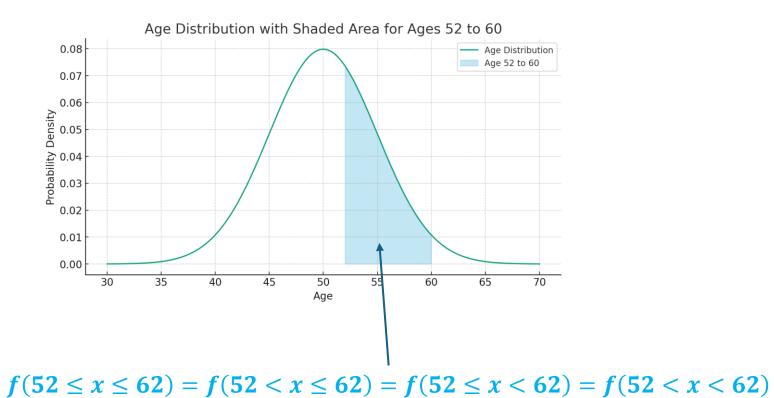
$$\int_{a}^{a} f(x) \ dx = 0$$



Example

- let's assume that in a certain population, the mean age is 50 years, and the standard deviation is 5 years.
- The shaded area represents the probability of selecting a person from this population whose age is at least 52 but not more than 60.





Probability density function

- For a continuous random variable X, the probability density function (PDF) is a function such that:
 - 1. $f(x_i) \ge 0$
 - $2. \quad \int_{-\infty}^{+\infty} f(x) \ dx = 1$
 - 3. $P(a \le x \le b) = \int_a^b f(x) dx$ = the area under the curve within the defined interval

Example

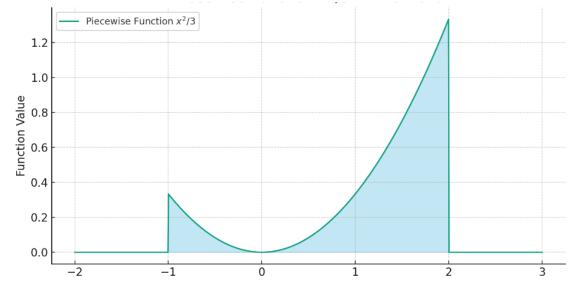
• let us assume that we have the following equation:

$$f(x) = \begin{cases} \frac{x^2}{3}, & -1 < x < 2, \\ otherwise, & 0 \end{cases}$$

- First, we make sure that this is an actual PDF, by calculating the area under this function within the given interval
 - This can be done by calculating the integral of this function for the defined interval

$$area = \int_{-1}^{2} \frac{x^2}{3} dx = \frac{x^3}{9} \Big|_{-1}^{2} = \frac{8}{9} - \frac{-1}{9} = 1$$

- Therefore, this can be considered as a pdf
- based on this you can calculate the probability of any interval within the function range using the same procedure
 - e.g., P(0.5 < x < 1)



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Cumulative distribution function

• For the PDF you can also calculate the CDF F(X).

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

This can help in calculating probabilities without performing integration

• given the CDF F(X), one can calculate the probability $P(a < X \le b)$ using:

$$F(X) = F(b) - F(a)$$

- Let us take an example of calculating the CDF for the previous function
 - and use it to find the probability $P(0 < X \le 1)$

$$f(x) = \begin{cases} \frac{x^2}{3}, & -1 < x < 2, \\ otherwise, & 0 \end{cases}$$

- We made sure that this is a PDF as its area is exactly 1
- Secondly, we find the integral for this function

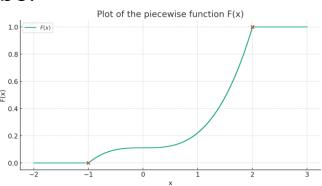
$$\int_{-1}^{+x} \frac{t^2}{3} dt$$

- note x becomes t, as x represents the limit of the integration (interval)
- we calculate the integral of this function to get

$$\frac{x^3}{9} - \frac{-1}{9} = \frac{x^3 + 1}{9}$$

• To make sure that x will not be out of its range, we rewrite it to be:

$$F(x) = \begin{cases} 0, & x < -1\\ \frac{x^3 + 1}{9}, & -1 \le x < 2\\ 1, & x \ge 2 \end{cases}$$



- now, to calculate P(0.5 < x < 1) we can use this function:
- $F(X) = F(b) F(a) = F(1) F(0.5) = \frac{2}{9} \frac{0.125}{9} = 0.0972$

Plot of the function F(x) with shaded area for P(0.5 < X < 1) 0.8 0.6 0.4 0.2 0.0 0.2 0.0

- Outside the range -1 and 2 the probability is 0 as we are subtracting the number from itself.
 - *P*(2.5<*X*<3) is 0

Exercise

- Find whether the following functions are a PDFs or not
- if yes, find its cumulative function

$$f(x) = \begin{cases} x^3, & 2 < x < 4, \\ otherwise, & 0 \end{cases}$$

$$g(x) = \begin{cases} 4x^3, & 0 < x < 1, \\ otherwise, & 0 \end{cases}$$