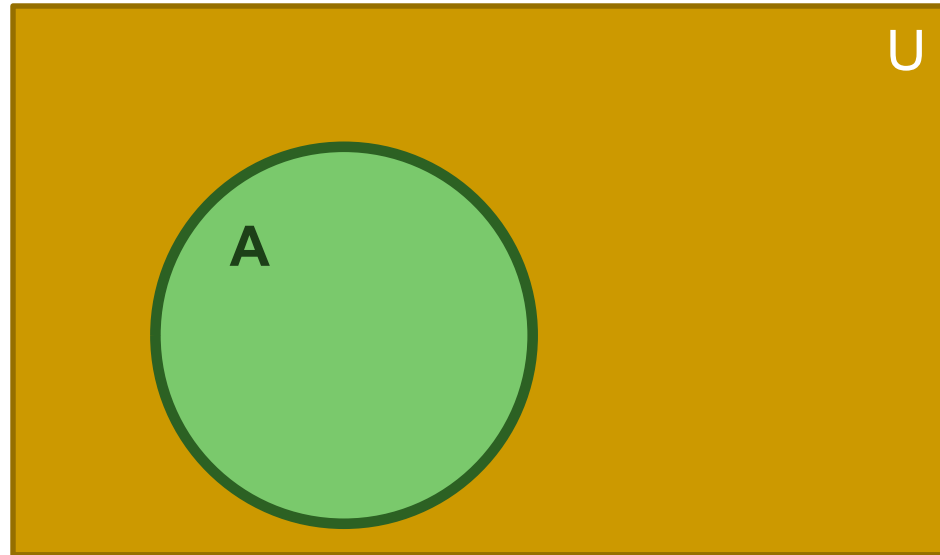

Probability Basics: A Gentle Introduction

Amitangshu Pal

Probability – Sample Space & Events

- Sample space is a collection of all possible outcomes in an experiment
- An event is a subset of the sample space



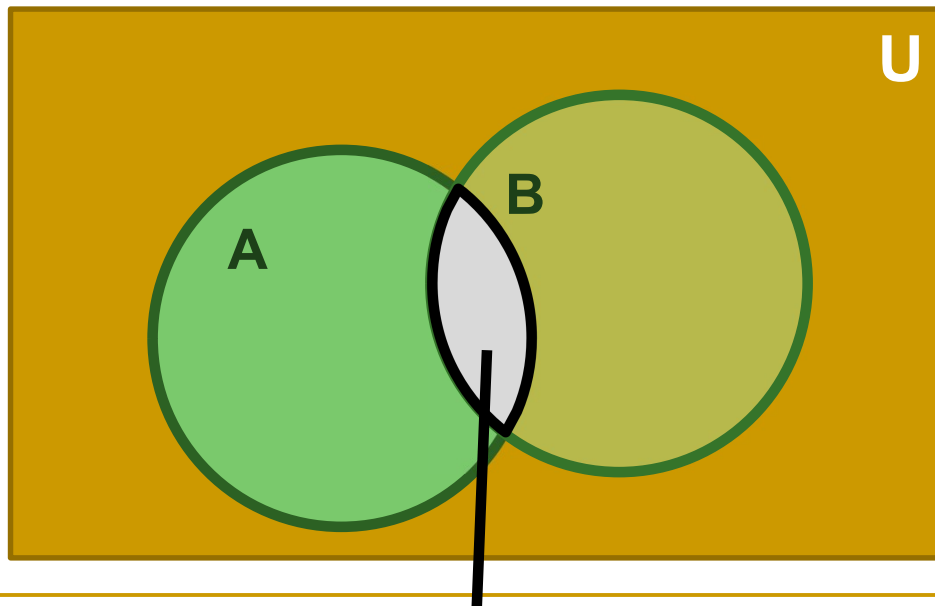
$$P[A] = \frac{|A|}{|U|}$$

Probability - Example

- Throw a fair coin thrice. What is the probability of getting exactly one tail?
 - All possible outcomes: $S = \{HHH, HHT, HTH, HTT, THH, THT, TTH, TTT\}$
 - Event of our interest is exactly one tail, i.e. $E = \{HHT, HTH, THH\}$
 - $P(E) = \frac{|E|}{|S|} = \frac{3}{8}$
-

Probability - Basics

- Consider the set of integers from 1 to 10
- Let A be the event of getting an even number $\rightarrow 2, 4, 6, 8, 10$
- Let B be the event of getting an integer greater than 5 $\rightarrow 6, 7, 8, 9, 10$
- What is the probability that a randomly drawn integer will be both even **and** greater than 5

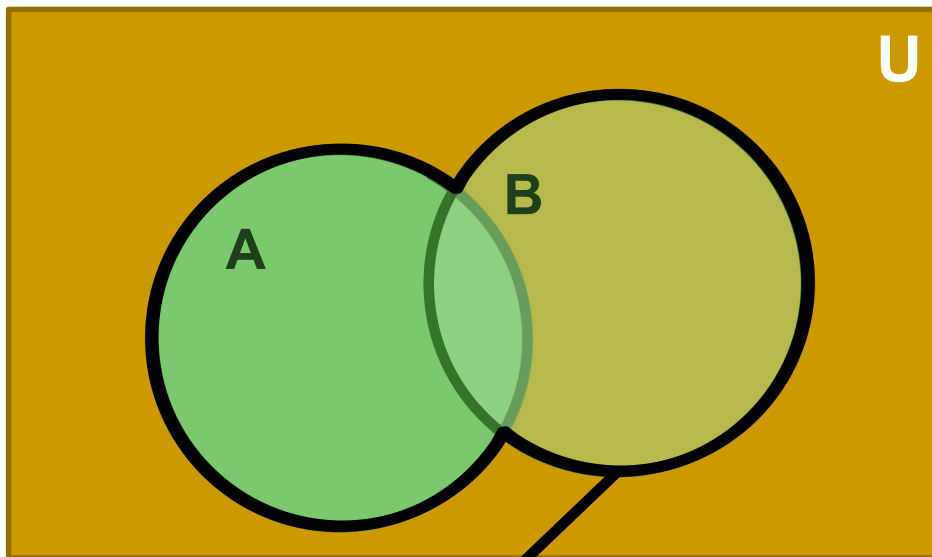


$$P[E] = \frac{|A \cap B|}{|U|} = \frac{3}{10}$$

$A \cap B$

Probability - Basics

- Consider the set of integers from 1 to 10
- Let A be the event of getting an even number $\rightarrow 2, 4, 6, 8, 10$
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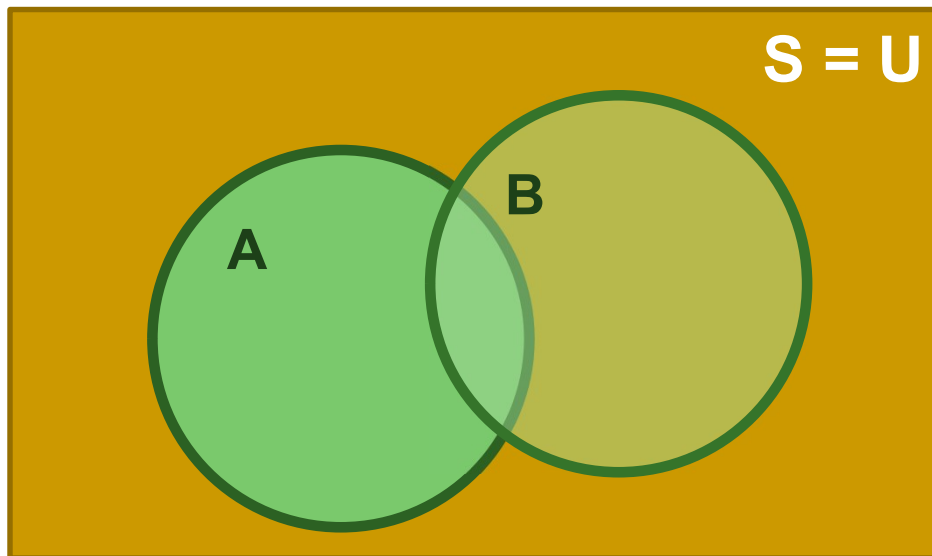
$A \cup B$

$$P[E] = \frac{|A \cup B|}{|U|}$$
$$= P[A] + P[B] - P[A \cap B]$$

Conditional Probability

- Suppose A has already occurred

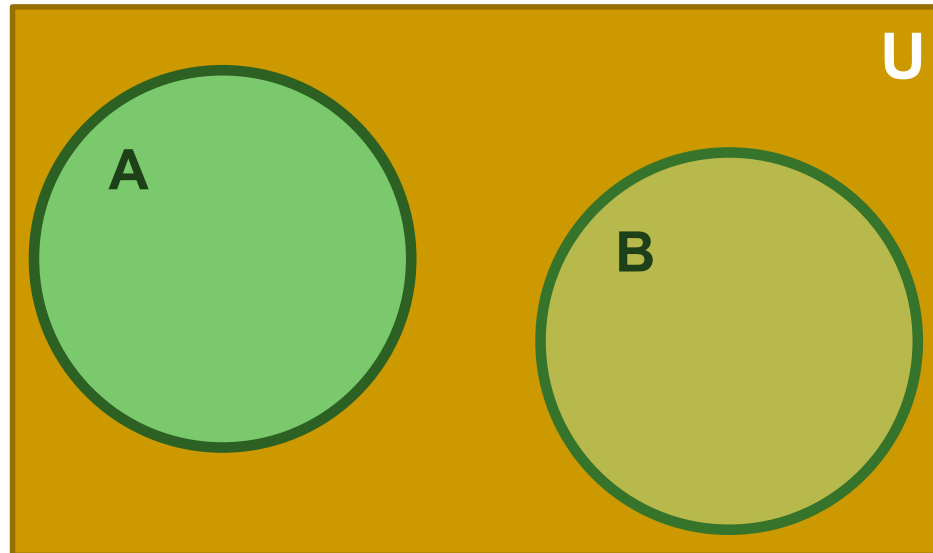
$$P[B|A] = \frac{\text{size of } A \cap B}{\text{size of } A} = \frac{\frac{\text{size of } A \cap B}{\text{size of } S}}{\frac{\text{size of } A}{\text{size of } S}} = \frac{P[A \cap B]}{P[A]}$$



$$P[A|B] = \frac{P[A \cap B]}{P[B]}$$

Independent Events

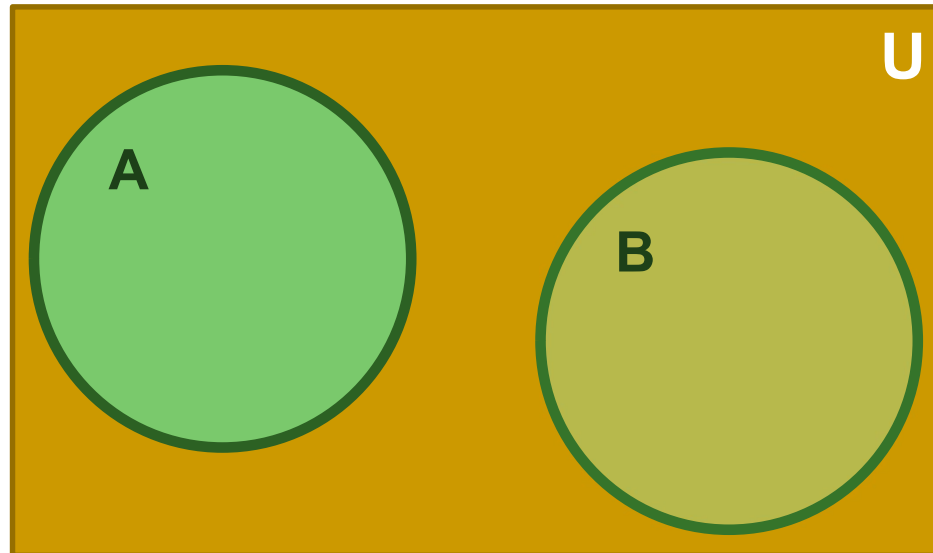
- Set A and B are independent if randomly picking from the set A tells us **nothing** about its membership in B



Are A and B independent?

Independent Events

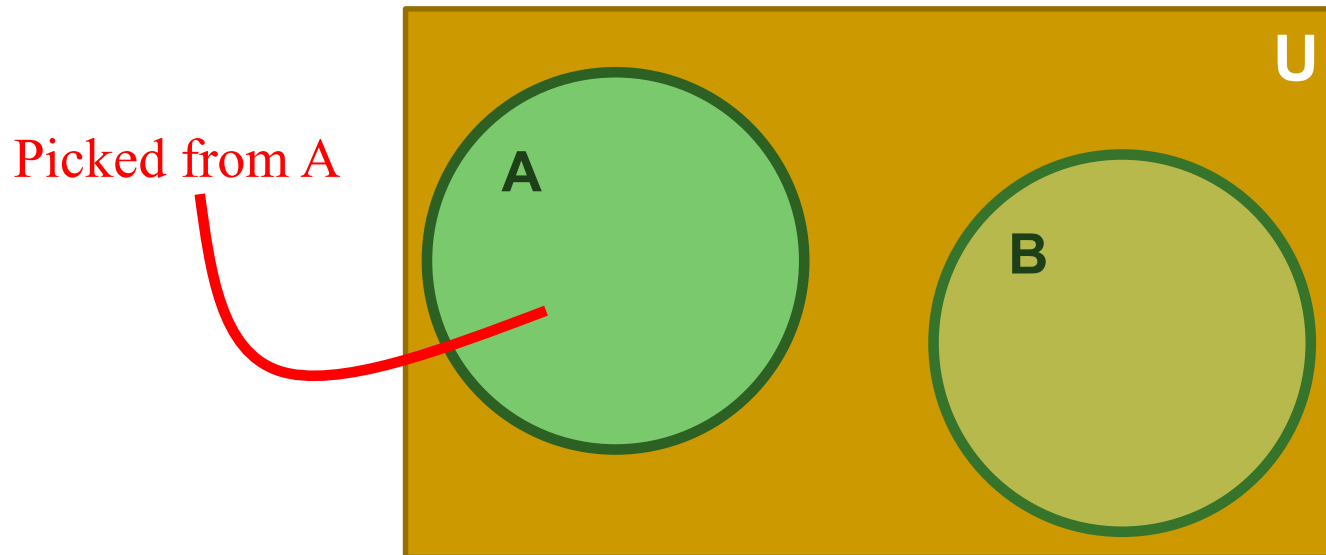
- Set A and B are independent if randomly picking from the set A tells us **nothing** about its membership in B



A and B are mutually exclusive

Independent Events

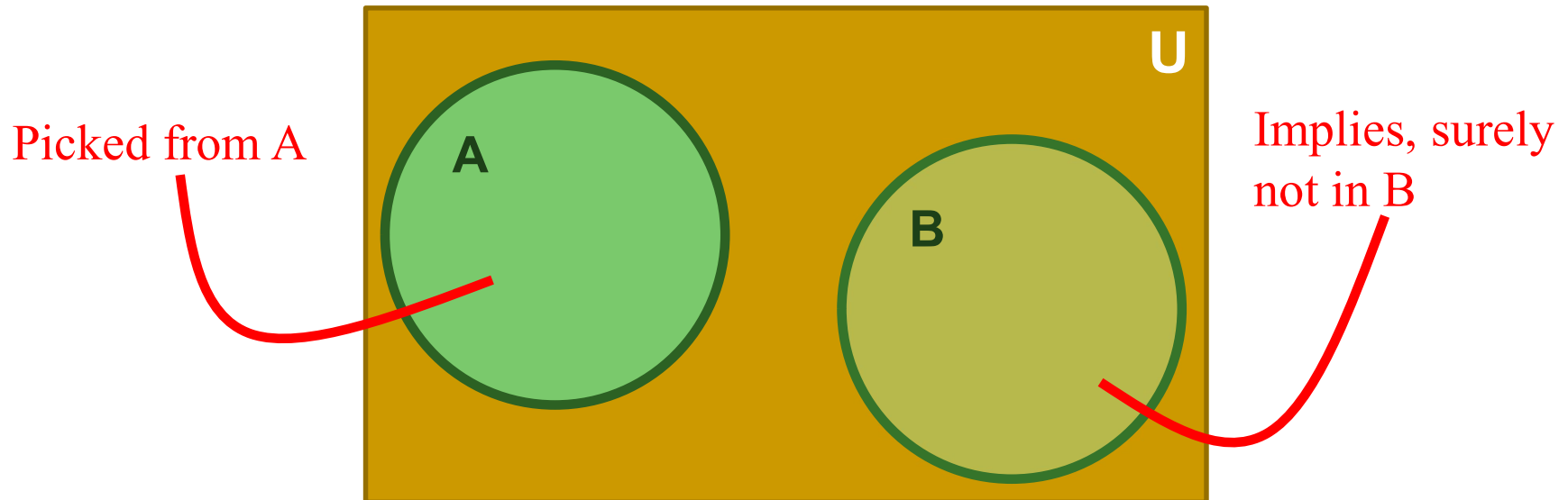
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A and B are mutually exclusive

Independent Events

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A and B are mutually exclusive

Independent Events

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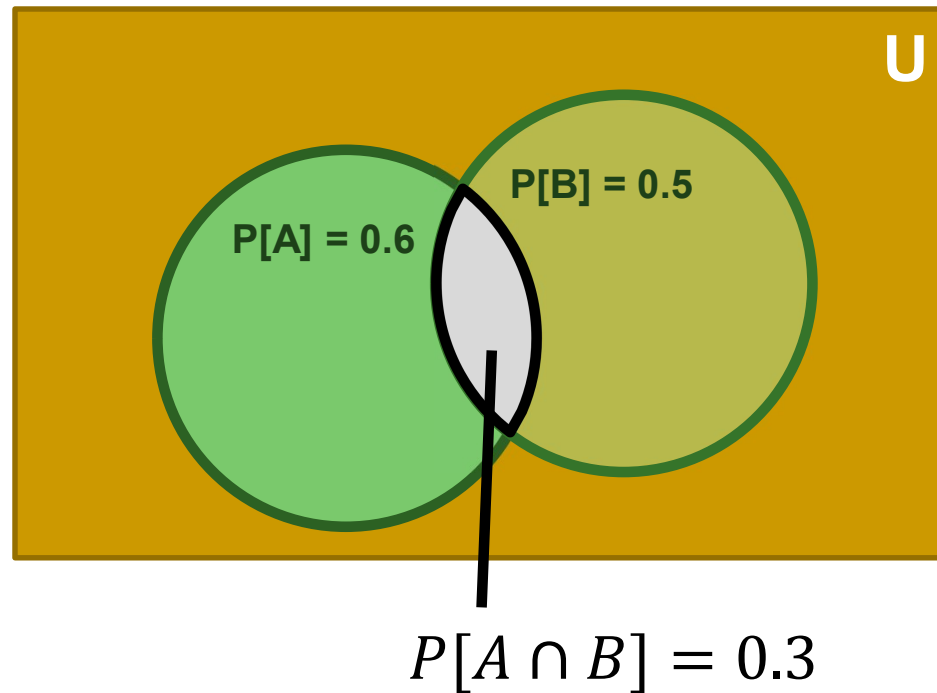
$$P[A|B] = P[A], \quad P[B|A] = P[B]$$

$$P[A|B] = \frac{P[A \cap B]}{P[B]}, \quad P[B|A] = \frac{P[A \cap B]}{P[A]}$$

$$P[A \cap B] = P[A|B] \cdot P[B]$$

$$P[A \cap B] = P[A] \cdot P[B]$$

Independent Events



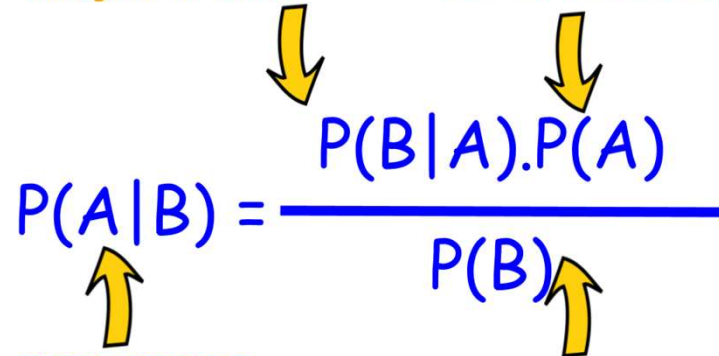
Bayes' theorem

LIKELIHOOD

The probability of "B" being True, given "A" is True

PRIOR

The probability "A" being True. This is the knowledge.

$$P(A|B) = \frac{P(B|A) \cdot P(A)}{P(B)}$$


POSTERIOR

The probability of "A" being True, given "B" is True

MARGINALIZATION

The probability "B" being True.

$$① \quad P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{P(B|A) \times P(A)}{P(B)}$$

$$② \quad P(B|A) = \frac{P(B \cap A)}{P(A)} = \frac{P(A \cap B)}{P(A)}$$

Bayes' theorem

10% of patients in a clinic have liver disease. Five percent of the clinic's patients are alcoholics. Amongst those patients diagnosed with liver disease, 7% are alcoholics. You are interested in knowing the probability of a patient having liver disease, given that he is an alcoholic.

$P(A)$ = probability of liver disease = 0.10

$P(B)$ = probability of alcoholism = 0.05

$P(B|A)$ = 0.07

$P(A|B)$ = ?

$$P(A|B) = \frac{P(B|A) \times P(A)}{P(B)} = \frac{0.07 \times 0.10}{0.05} = 0.14$$

In other words, if the patient is an alcoholic, their chances of having liver disease is 0.14 (14%)

Random Variable

Definition: Random Variable

A random variable is a rule that assigns a numerical value to an outcome of interest.

Example: Suppose we toss a coin 3 times, we define a random variable X as the number of outcomes where a tail appears.

This random variable can take values of 0, 1, 2 *and* 3.

- The probability that the random variable takes a given value can be computed using the rules governing probability.
 - For example, the probability that $X = 1$ means there will be 1 tail. Symbolically, it is denoted as $P(X=1) = 3/8$

Probability Distribution

Definition: **Probability distribution**

A probability distribution is a definition of probabilities of the values of random variable.

Example: Then the probability distribution is given by

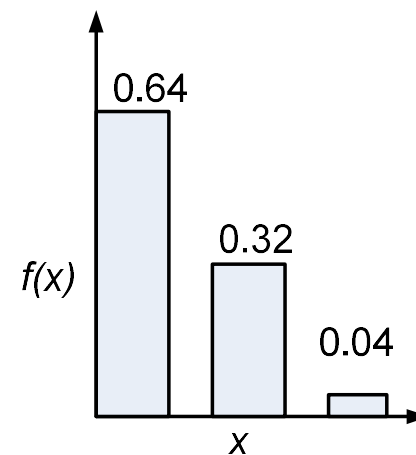
X	Probability
0	1/8
1	3/8
2	3/8
3	1/8

Probability Distribution

- In general, a probability distribution function takes the following form

x	x_1	$x_2 \dots \dots \dots x_n$
$f(x) = P(X = x)$	$f(x_1)$	$f(x_2) \dots \dots \dots f(x_n)$

x	0	1	2
$f(x)$	0.64	0.32	0.04



Descriptive measures

Given a random variable X in an experiment, we have denoted $f(x) = P(X = x)$, the probability that $X = x$. For discrete events $f(x) = 0$ for all values of x except $x = 0, 1, 2, \dots$

Properties of discrete probability distribution

$$0 \leq f(x) \leq 1$$

$$\sum f(x) = 1$$

$$\mu = \sum x \cdot f(x)$$

[is the **mean**]

$$\sigma^2 = \sum (x - \mu)^2 \cdot f(x)$$

[is the **variance**]

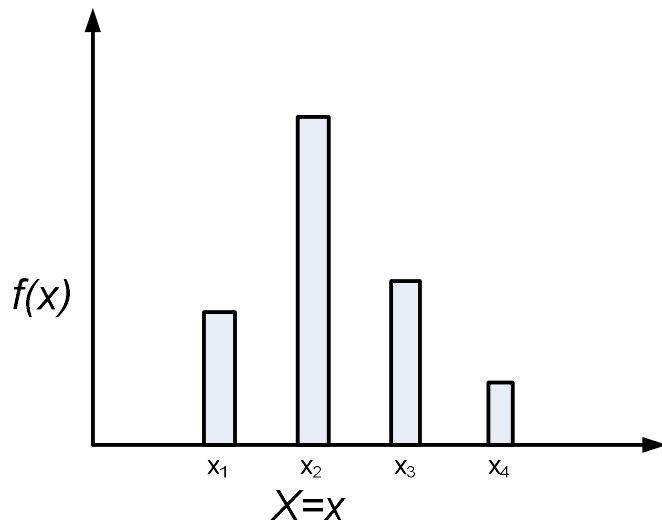
Note: For discrete **uniform** distribution, $f(x) = \frac{1}{n}$ with $x = 1, 2, \dots, n$

$$\mu = \frac{1}{n} \sum_{i=1}^n x_i$$

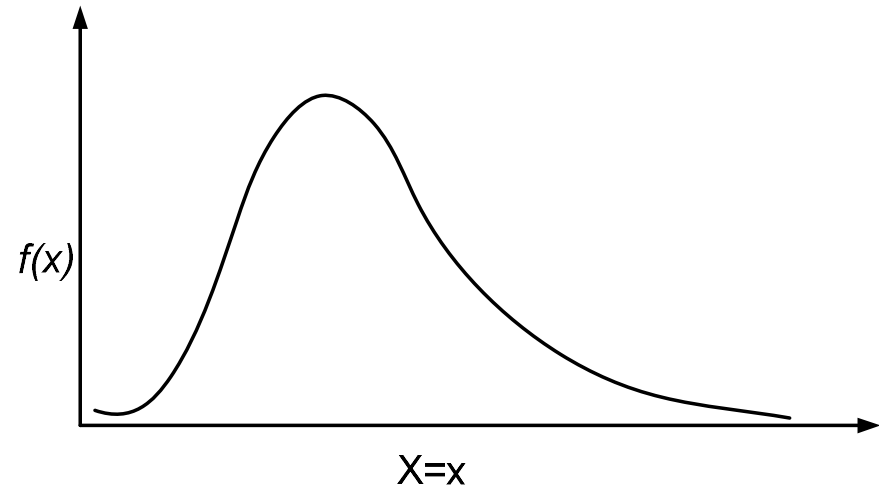
$$\text{and } \sigma^2 = \frac{1}{n} \sum_{i=1}^n (x_i - \mu)^2$$

Continuous Probability Distributions

- When the random variable of interest can take **any value in an interval**, it is called continuous random variable.
- Every continuous random variable has **an infinite, uncountable number of possible values** (i.e., any value in an interval)



Discrete Probability distribution



Continuous Probability Distribution

Properties of Probability Density Function

The function $f(x)$ is a probability density function for the continuous random variable X , defined over the set of real numbers R , if

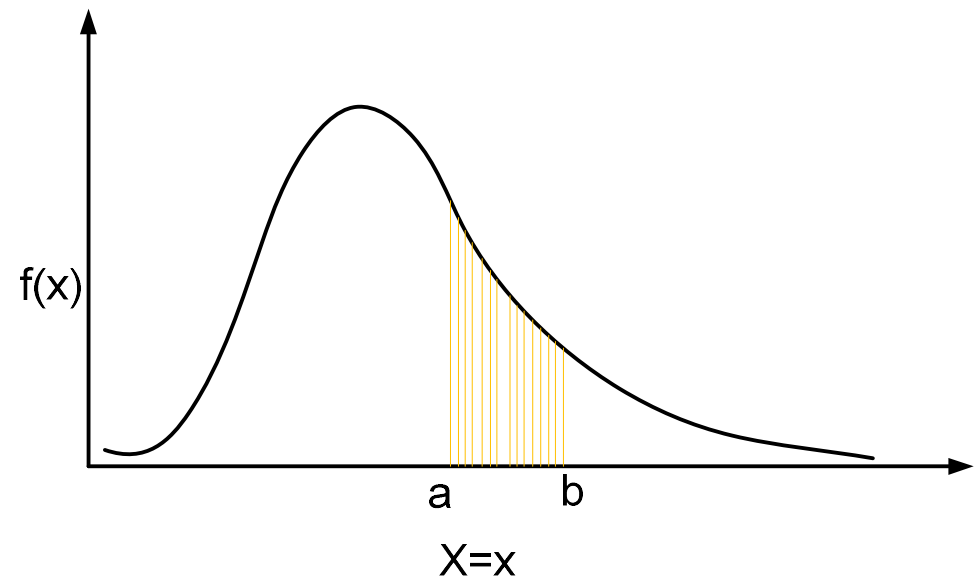
$$f(x) \geq 0, \text{ for all } x \in R$$

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

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

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

$$\sigma^2 = \int_{-\infty}^{\infty} (x - \mu)^2 f(x) dx$$



Continuous Uniform Distribution

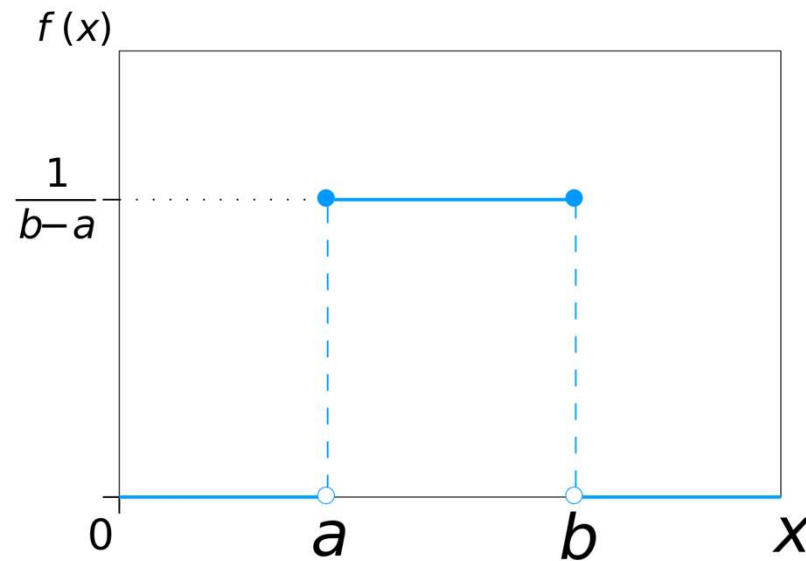
- One of the simplest continuous distribution in all of statistics is the continuous **uniform** distribution.

Definition: Continuous Uniform Distribution

The density function of the continuous uniform random variable X on the interval $[a, b]$ is:

$$f(x; a, b) = \begin{cases} \frac{1}{b - a} & a \leq x \leq b \\ 0 & \text{Otherwise} \end{cases}$$

Continuous Uniform Distribution



Note:

a) $\int_{-\infty}^{\infty} f(x)dx = \frac{1}{b-a} \times (b-a) = 1$

b) $P(c < x < d) = \frac{d-c}{b-a}$ where both c and d are in the interval (a,b)

c) $\mu = \frac{a+b}{2}$

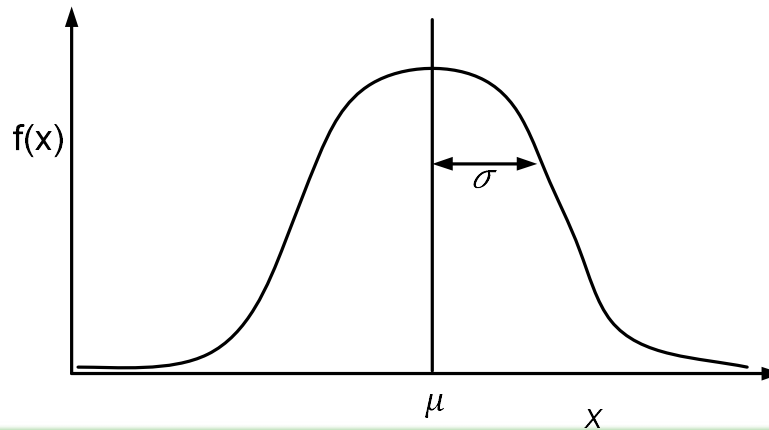
d) $\sigma^2 = \frac{(b-a)^2}{12}$

Normal Distribution

- The most often used continuous probability distribution is the normal distribution; it is also known as **Gaussian distribution**.
 - Its graph called the normal curve is the bell-shaped curve.
 -
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Normal Distribution

- The most often used continuous probability distribution is the normal distribution; it is also known as **Gaussian distribution**.
- Its graph called the normal curve is the bell-shaped curve.

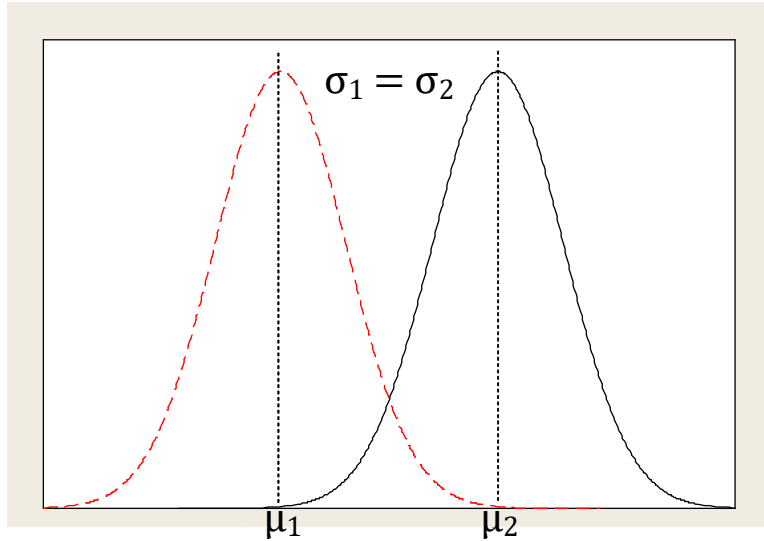


Definition: Normal distribution

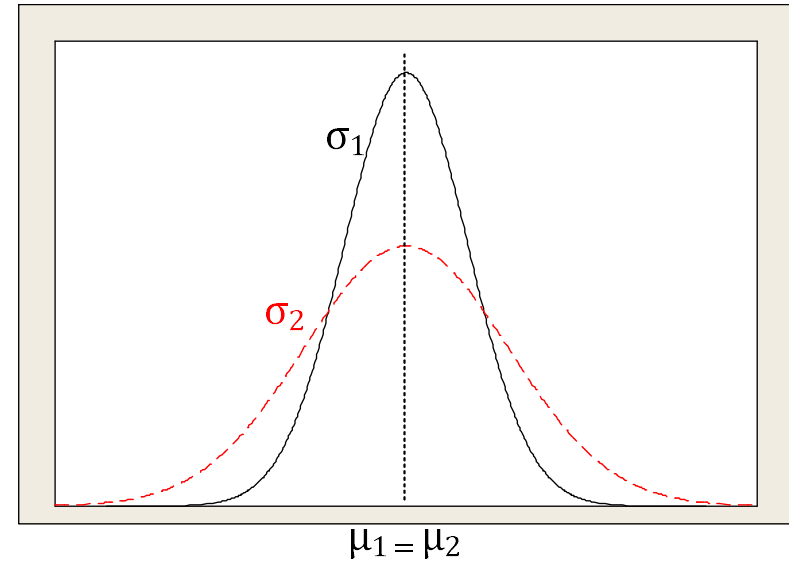
The density of the normal variable x with mean μ and variance σ^2 is

$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} \quad -\infty < x < \infty$$

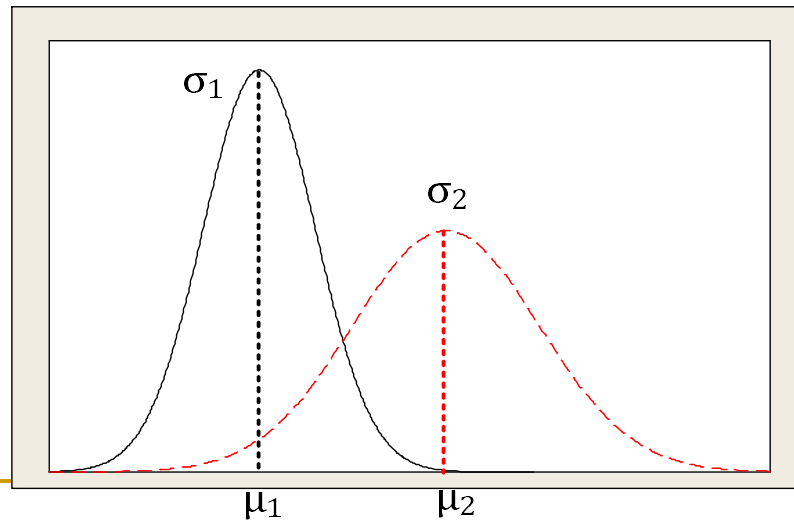
Normal Distribution



Normal curves with $\mu_1 < \mu_2$ and $\sigma_1 = \sigma_2$



Normal curves with $\mu_1 = \mu_2$ and $\sigma_1 < \sigma_2$



Normal curves with $\mu_1 < \mu_2$ and $\sigma_1 < \sigma_2$

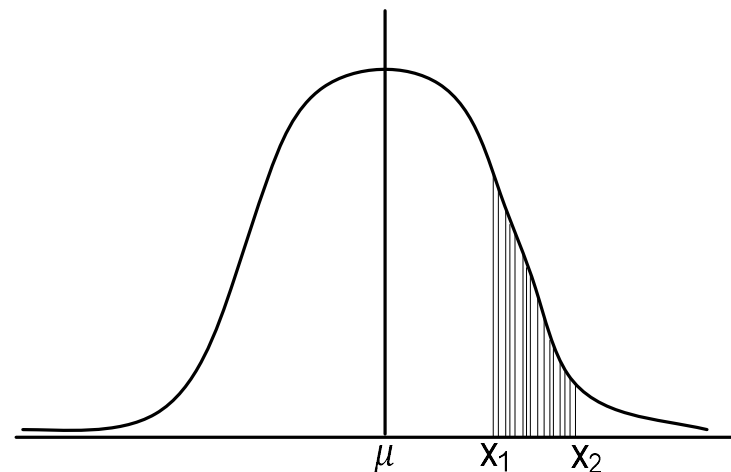
Properties of Normal Distribution

$$\int_{-\infty}^{\infty} f(x)dx = \frac{1}{\sigma\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-\frac{1}{2\sigma^2}(x-\mu)^2} dx = 1$$

$$\mu = \int_{-\infty}^{\infty} x \cdot f(x)dx = \frac{1}{\sigma\sqrt{2\pi}} \int_{-\infty}^{\infty} x \cdot e^{-\frac{1}{2\sigma^2}(x-\mu)^2} dx$$

$$\sigma^2 = \frac{1}{\sigma\sqrt{2\pi}} \int_{-\infty}^{\infty} (x - \mu)^2 \cdot e^{-\frac{1}{2\sigma^2}(x-\mu)^2} dx$$

$$P(x_1 < x < x_2) = \frac{1}{\sigma\sqrt{2\pi}} \int_{x_1}^{x_2} e^{-\frac{1}{2\sigma^2}(x-\mu)^2} dx$$



Examples of Random Variables

Continuous-valued random variable:

Uniform:

$$p(x) = 1 \quad 0 \leq x \leq 1$$

Discrete-valued random variable:

$$p(x_i) = \frac{1}{M} \quad i = 0, \dots, M - 1$$

Gaussian:

$$p(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}} \quad -\infty \leq x \leq \infty$$

N/A

Check: <http://mathlets.org/mathlets/probability-distributions/>
