



Preliminaries: Probability

CS 418. Introduction to Data Science

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Preliminaries Terminology

- An **experiment** is a procedure that results in one of a number of possible **outcomes**.
 - *Example:* rolling a dice.
- The **sample space** of the experiment is the set of all possible **outcomes**.
 - *Example:* $\{1, 2, 3, 4, 5, 6\}$



- An **event** is a subset of the sample space.
 - *Example:* $\{2, 4, 6\}$





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Probability of an Event

- **Probability** is the measure of the **likelihood** of an event occurring.

If S is a finite nonempty sample space of **equally likely outcomes**, and $E \subseteq S$ is an **event**, then the **probability** of E is

$$P(E) = \frac{|E|}{|S|}$$

- The **probability** of an event is **between 0 and 1**.
- **Example:** Two dice are rolled.

What is the probability that the numbers on the two dice are the same?

$$P(E) = \frac{6}{36} = \frac{1}{6}$$



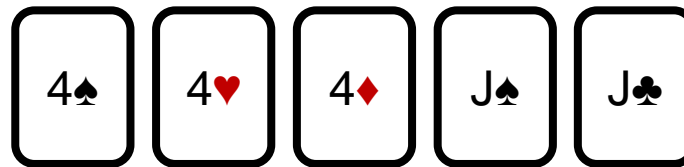
Preliminaries

Exercise 3.1



**THINK
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What is the probability that a hand of 5 cards drawn from a standard deck of 52 cards contains a full house (3 cards of one rank and 2 of another rank)?



$$\frac{\binom{13}{1}\binom{4}{3}\binom{12}{1}\binom{4}{2}}{\binom{52}{5}}$$

Each card has a **rank** and a **suit**. There are **13 possible ranks**, **4 possible suits**, and **13 cards for each suit** (one for each rank).



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Complement of an Event

- The **complement** of an event E , denoted E' , consists of all outcomes that are **not** in E .

Let E be an event in a sample space S .
The **probability of the complement** of E is

$$P(E') = 1 - P(E)$$

- Example:** Two dice are rolled.

What is the probability that the numbers on the two dice are not the same?

$$P(E') = 1 - \frac{1}{6} = \frac{5}{6}$$



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Mutually Exclusive Events

- The **joint probability** of E and F , denoted $P(E \cap F)$, is the probability of **both** E and F occurring at the same time.
- Two events are **mutually exclusive** if they cannot occur at the same time.

The events E and F are **mutually exclusive** if and only if

- **Example:** Two dice are rolled. E is the event that the numbers on the two dice are the same, F is the event that the number on the first die is even, and G is the event that the number on the first die is odd.

Are E and F mutually exclusive? No

Are F and G mutually exclusive? Yes



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Conditional Probability

- The **conditional probability** of E given F , denoted $P(E|F)$, is the probability of E occurring **given that** F has already occurred.

Let E and F be events with $P(F) > 0$. The **conditional probability of E given F** , denoted $P(E|F)$, is given by

$$P(E|F) = \frac{P(E \cap F)}{P(F)}$$

- Example:** Two dice are rolled.

What is the probability that the numbers on the two dice are the same given that the number on the first die is even?

$$P(E|F) = \frac{3/36}{1/2} = \frac{1}{6}$$



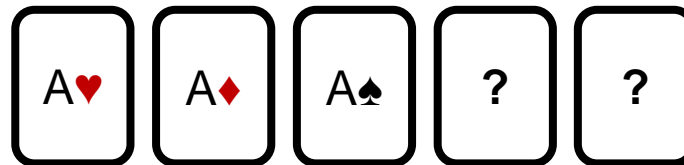
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Exercise 3.2



**THINK
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What is the probability that a hand of 5 cards drawn from a standard deck of 52 cards contains a full house given that the first three cards drawn are the ace of hearts, the ace of diamonds, and the ace of spades?



$$\frac{\binom{12}{1} \binom{4}{2}}{\binom{49}{2}}$$

Conditional probability

$$P(E|F) = \frac{P(E \cap F)}{P(F)}$$

Each card has a **rank** and a **suit**. There are **13 possible ranks**, **4 possible suits**, and **13 cards for each suit** (one for each rank).

Preliminaries Independent Events

- Two events are **independent** if knowing that one occurred does **not** affect the probability of the other.

The events **E** and **F** are **independent** if and only if

$$P(E|F) = P(E)$$

$$P(F|E) = P(F)$$

$$P(E \cap F) = P(E) \cdot P(F)$$

All these statements
are **equivalent**

- Example:** Two dice are rolled. **E** is the event that the numbers on the two dice are the same and **F** is the event that the number on the first die is even.

Are E and F independent? $\forall \epsilon$

Preliminaries Bayes' Theorem (I)

- Conditional probabilities can be correctly “reversed” using **Bayes' theorem**.

Let **A** and **B** be events with **$p(A) \neq 0$** and **$p(B) \neq 0$** . Then

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

- Example:** Suppose that 1 in 100,000 people has a rare disease. 99.0% of the people who have the disease test positive and 99.5% of the people who do not have the disease test negative. **What is the probability that a person who tests positive has the disease?**

$$P(\text{disease}|+) = \frac{P(+|\text{disease}) P(\text{disease})}{P(+|\text{disease}) P(\text{disease}) + P(+|\neg \text{disease}) P(\neg \text{disease})} \approx 0.002$$



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Bayes' Theorem (II)

Let A and B be events with $p(A) \neq 0$ and $p(B) \neq 0$. Then

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

- **Example:** Suppose that we have found that the word “free” occurs in 250 out of 2000 emails known to be spam and in 5 out of 1000 emails known not to be spam. Assume that it is equally likely that an incoming message is spam or not spam. **What is the probability that an incoming message containing the word “free” is spam?**

$$P(\text{spam}|\text{free}) = \frac{P(\text{free}|\text{spam}) P(\text{spam})}{P(\text{free}|\text{spam}) P(\text{spam}) + P(\text{free}|\neg\text{spam}) P(\neg\text{spam})} \approx 0.962$$



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Random Variables

- A **random variable** X is a function from the sample space of an experiment to the set of real numbers.
- The **range of a random variable** X is the set of all possible values of X .
- A **random variable** can be **discrete** (if its range is countable) or **continuous** (if its range is uncountable).
- **Example:**

- Let the **random variable** X be the **number of heads that come up** when a coin is flipped 3 times.

What are the possible values of X ?

$\{0, 1, 2, 3\}$

- Let the **random variable** Y be the **time used by a student** to complete a timed 60-minute exam.

What are the possible values of Y ?

$[0, 60]$



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Discrete Random Variables

- The probability distribution of a **discrete** random variable X can be described by a **probability mass function (pmf)**.
- A **probability mass function** $P(X)$ assigns a probability to each possible value of X .
 - Each probability is **between zero and one**, inclusive.
 - The **sum of the probabilities is one**.
- **Example:**
 - Let the **random variable** X be the **number of heads that come up** when a coin is flipped 3 times.

What is the probability mass function of X ?

$$P(X = 0) = 1/8$$

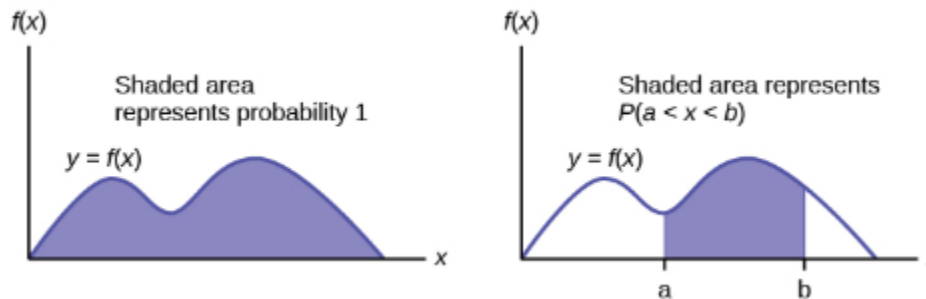
$$P(X = 1) = 3/8$$

$$P(X = 2) = 3/8$$

$$P(X = 3) = 1/8$$

Preliminaries Continuous Random Variables

- The probability distribution of a **continuous** random variable can be described by a **probability density function (pdf)**.
- A **probability mass function $f(x)$** describes the relative likelihood of all possible values of X .
 - **$f(x) \geq 0$.**
 - The **total area under the curve $f(x)$ is one.**



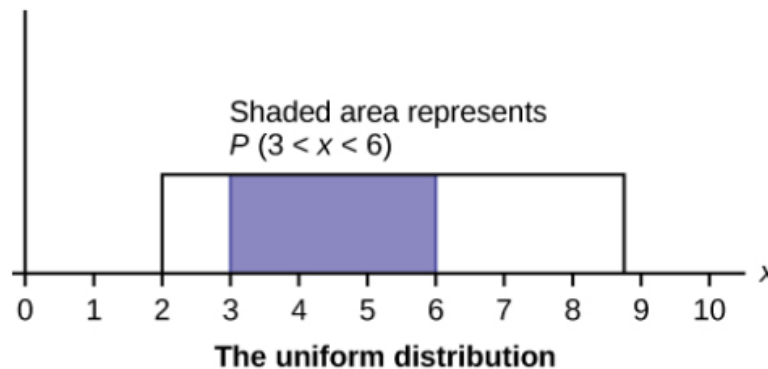
Source: OpenStax, [Introductory Statistics](#) (2016)



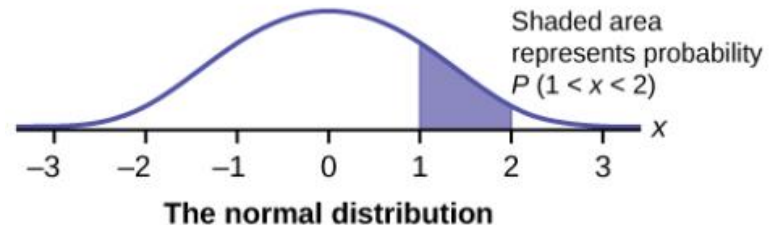
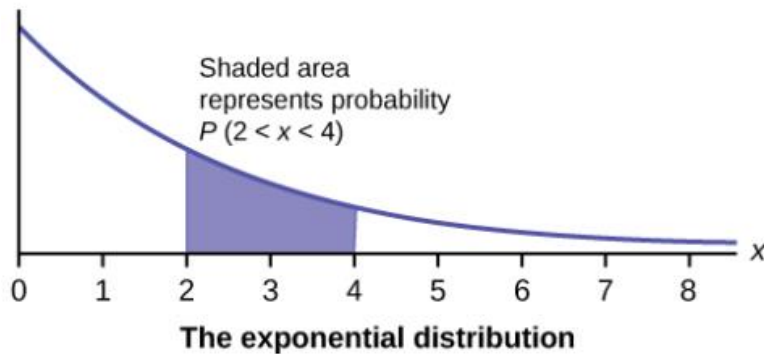
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Continuous Probability Distribution

- There are many **continuous probability distributions**.



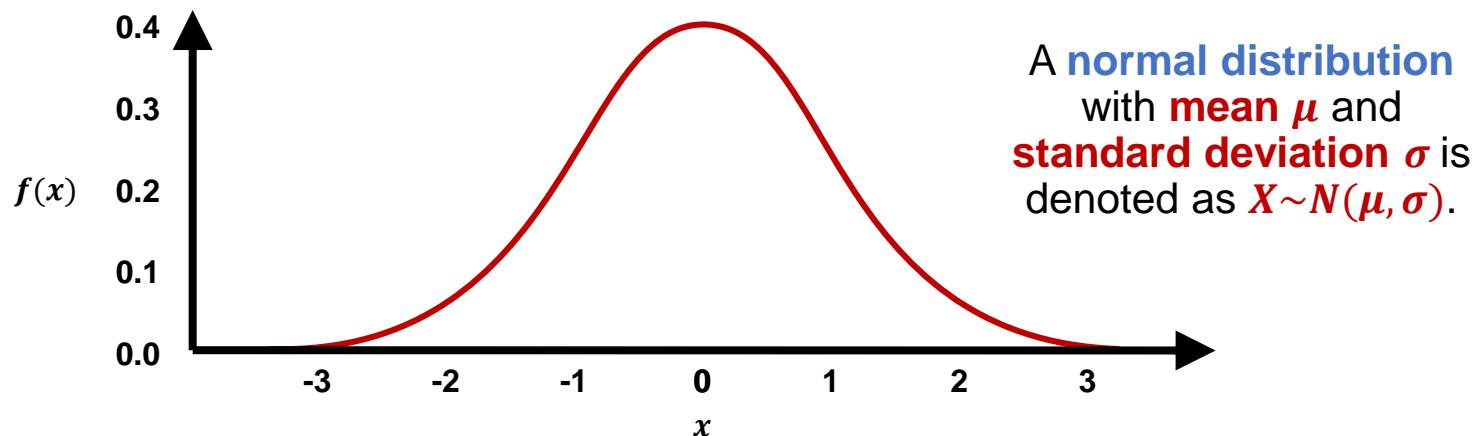
The **uniform distribution** is concerned with **outcomes that are equally likely to occur**



Source: OpenStax, [Introductory Statistics](#) (2016)

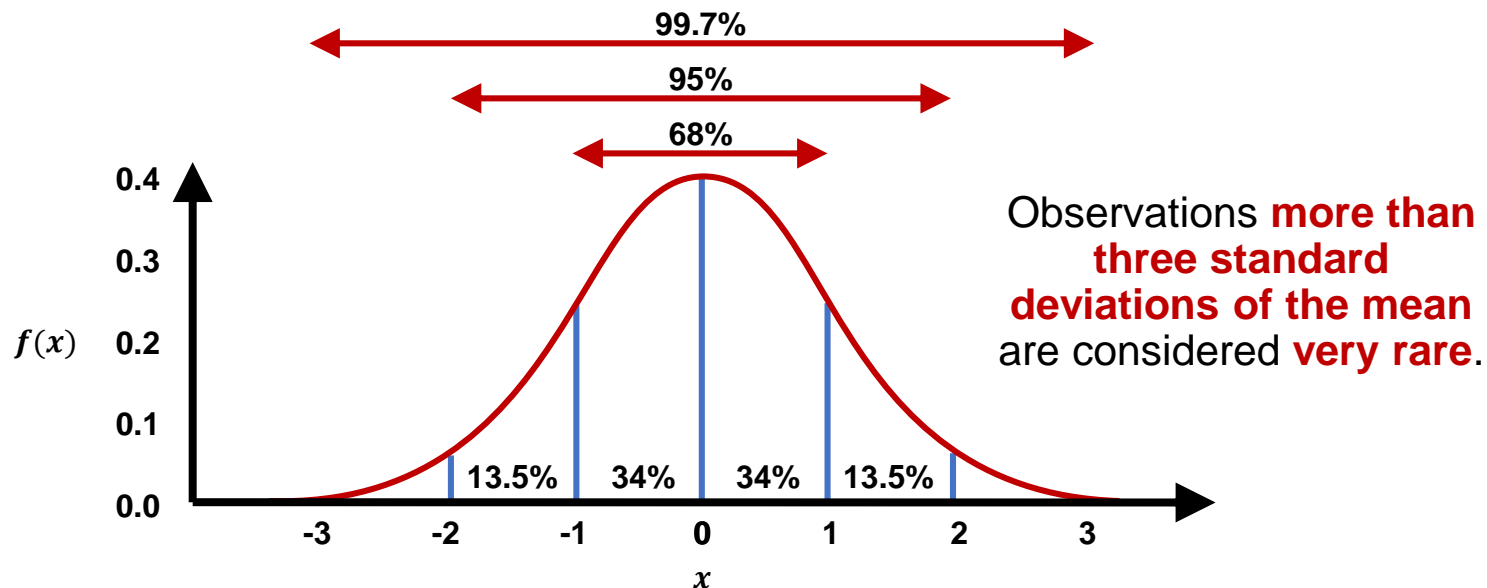
Preliminaries Normal Distribution

- The **normal (Gaussian) distribution** is a **continuous** distribution characterized by a **symmetric, unimodal, bell-shaped** curve.
- The **normal distribution** has two **parameters**: the **mean μ** and the **standard deviation σ** .
- Many processes can be approximated by the **normal distribution**, including exam scores and heights or other physical measurements.



Normal Distribution The Empirical Rule

- The **empirical rule** (also known as the **68-95-99.7 rule**) states that, for any **normal distribution**:
 - **68%** of the data falls within **one standard deviation** of the mean.
 - **95%** of the data falls within **two standard deviations** of the mean.
 - **99.7%** of the data falls within **three standard deviations** of the mean.





Preliminaries References

- David Diez, Christopher Barr, and Mine Çetinkaya-Rundel. [*OpenIntro Statistics*](#) (2015).
- Joel Grus. *Data Science from Scratch* (2015).
- OpenStax. [*Introductory Statistics*](#) (2016).