

# Statistics Foundations: Understanding Probability and Distributions

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## INTRODUCING THE CONCEPT OF PROBABILITY



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# Statistics

A branch of mathematics dealing with the collection, analysis, interpretation, presentation, and organization of data.



# How Statistics Works

## **Formulate a hypothesis**

- “Smoking causes cancer”

## **Make observations**

- Get data regarding smoking habits and medical history

## **Analyze data and make conclusions**

- Accept or reject hypothesis
- Modify experiment, get more data, etc.



# Course Overview



1. **Introducing the Concept of Probability**
2. **Calculating the Conditional Probability of Events**
3. **Understanding Random Variables and Distributions**
4. **Introducing the Concept of Expectation**
5. **Looking at Some Special Statistical Distributions**



# Structure

**First course in the Statistics Foundations series**

**Theory lectures**

**Live examples in R**

**Simulation/analysis comparison**



Goal:

Understand the concept of *probability*; learn the basics of set theory and combinatorics.



# Overview



Naïve Set Theory Primer

Experiments and Events

Probability

Counting Methods

Combinatorics

Probability of a Union of Events



# Naïve Set Theory Primer

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# Set Theory

A branch of mathematics that deals with sets.

Naïve Set Theory = set theory explained in simple terms, without resorting to heavy math.

(Alternative: Axiomatic Set Theory)



# Set

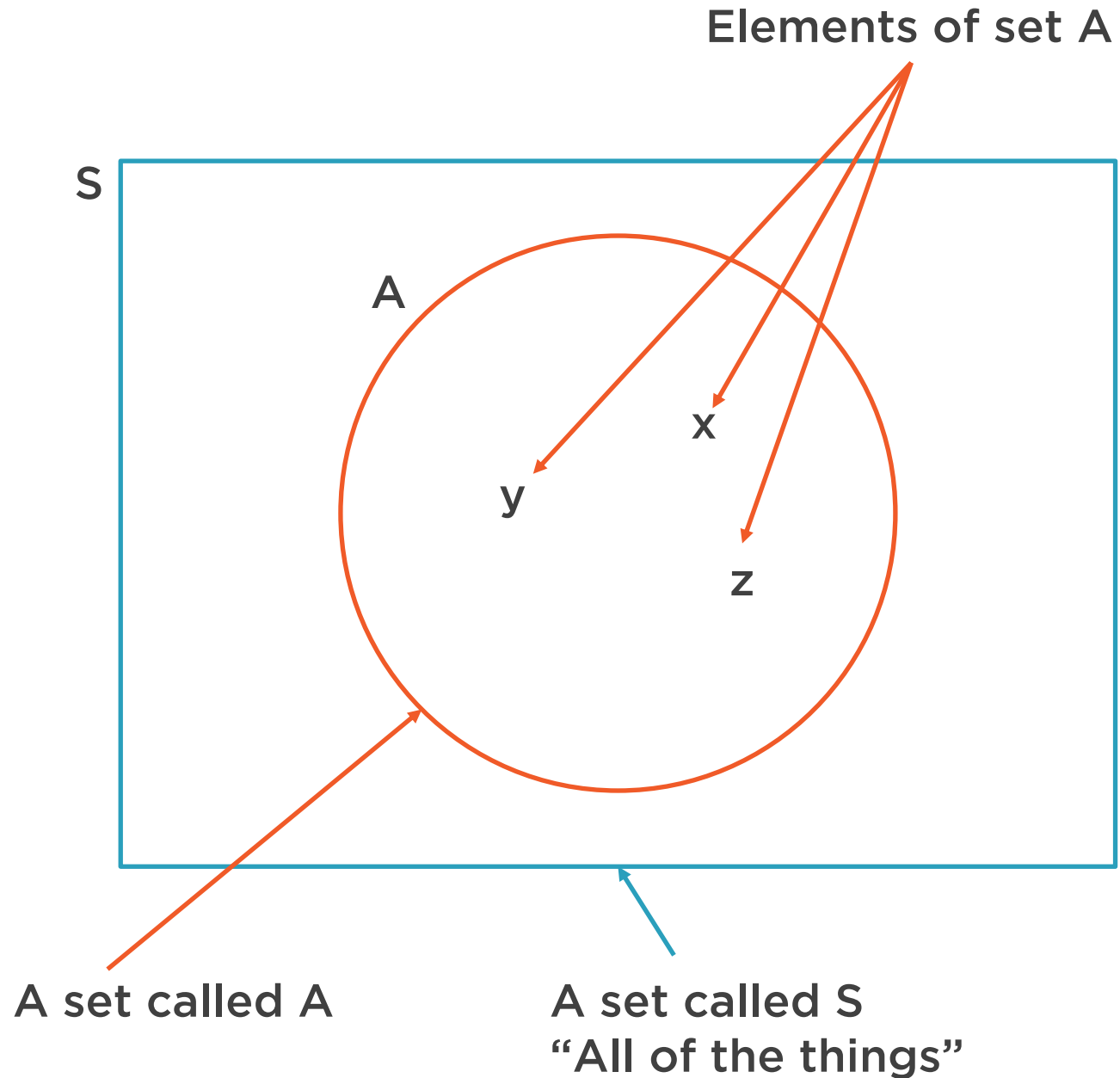
A well-defined collection of distinct objects.

$A = \{0, 1, 2, 3, 4, 5, 6, 7, 8, 9\}$  a set of digits

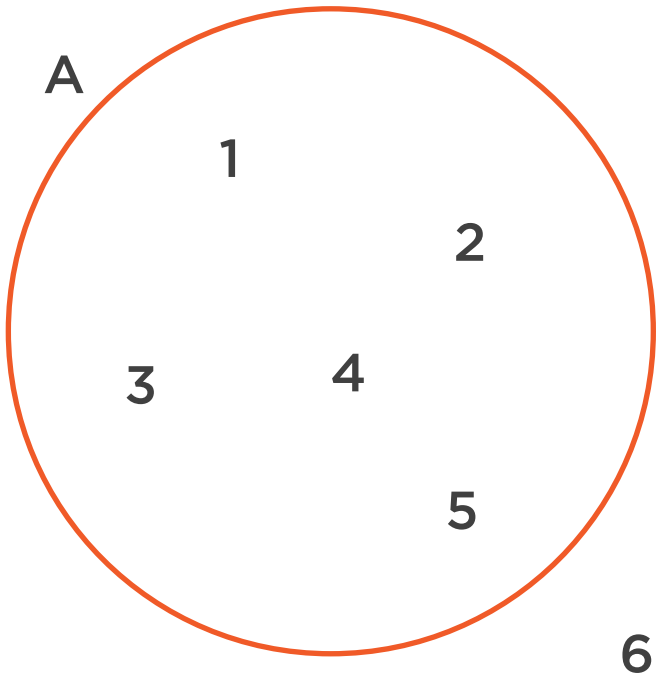
$B = \{a, b, c, \dots, z\}$  a set of letters

$S = \{\boxplus, \ominus, *, \wedge\}$  a set of weird mathematical operators

Sets are often  
illustrated with Venn  
diagrams  
Used for illustration in  
this course



# Set Membership



Consider a set  $A = \{1, 2, 3, 4, 5\}$

We can state that 1 is an element of  $A$

$$1 \in A$$

And that 6 is *not* an element of  $A$

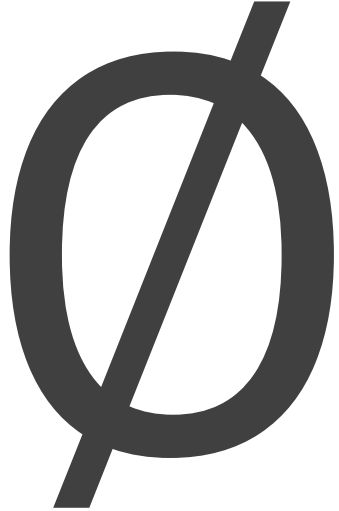
$$6 \notin A$$

This notation is often used to specify a set of values

E.g., 'for all even natural numbers':

$$\forall x : \{x \in \mathbb{N} \mid x \text{ is even}\}$$

# Empty Set



**A set containing no elements**

$$\emptyset = \{\}$$

**For every element  $x$ ,  $x \notin \emptyset$**

# Subsets

Set  $A$  is a subset of set  $B$  if every element of  $A$  is in  $B$

For example, given  $A = \{1,2,3\}$ ,  $B = \{1,2,3,4,5\}$ ,  $A$  is a subset of  $B$ :

$$A \subset B$$

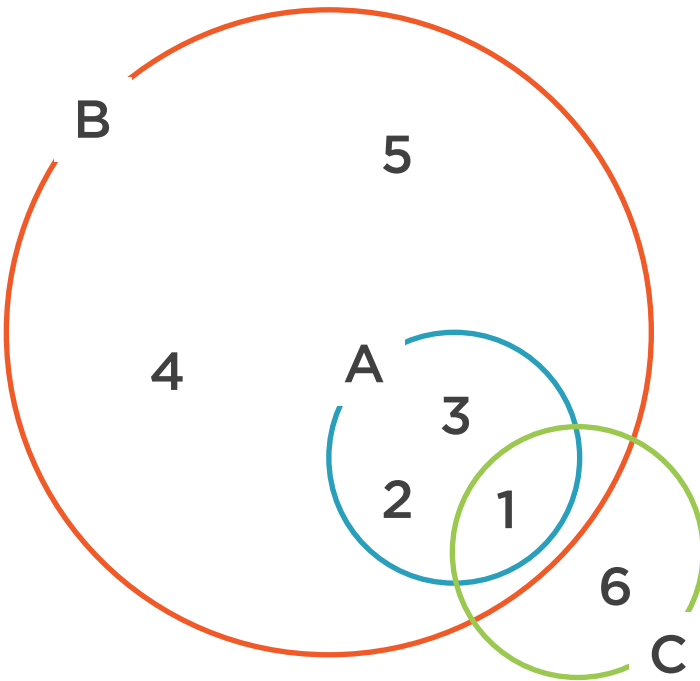
Similarly, we have a 'not a subset of' operator, so if  $C = \{1,6\}$

$$C \not\subset B$$

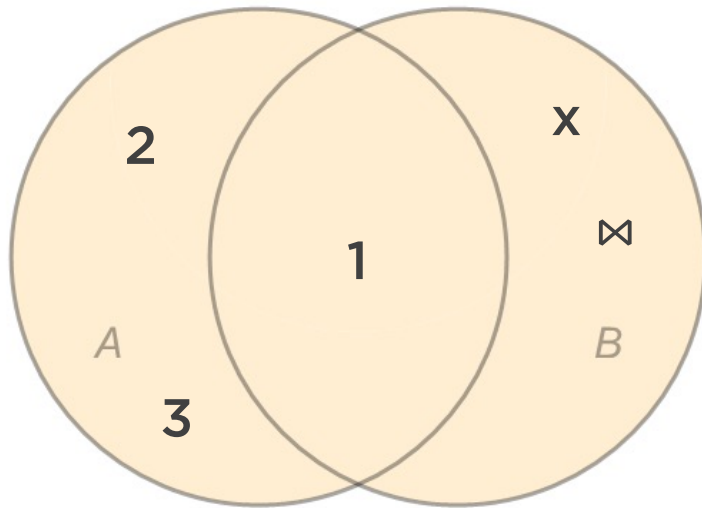
For every set  $A$

$$\emptyset \subset A$$

$$A \subset A$$



# Set Union



**A union of two sets  $A$  and  $B$  is a set containing all elements from  $A$  and  $B$**

**E.g., if  $A = \{1, 2, 3\}$  and  $B = \{1, x, \infty\}$  then**

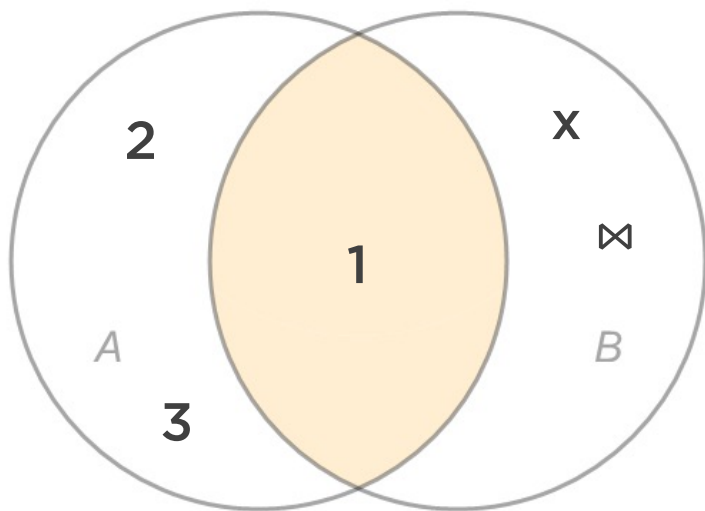
$$A \cup B = \{1, 2, 3, x, \infty\}$$

**Notice that the common elements do not repeat**

**Also exists as a large operator**

$$\bigcup_{i=1}^3 A_i = A_1 \cup A_2 \cup A_3$$

# Set Intersection



**An *intersection* of two sets  $A$  and  $B$  is a set containing all that belong to both  $A$  and  $B$**

**E.g., if  $A = \{1, 2, 3\}$  and  $B = \{1, x, \infty\}$  then**

$$A \cap B = \{1\}$$

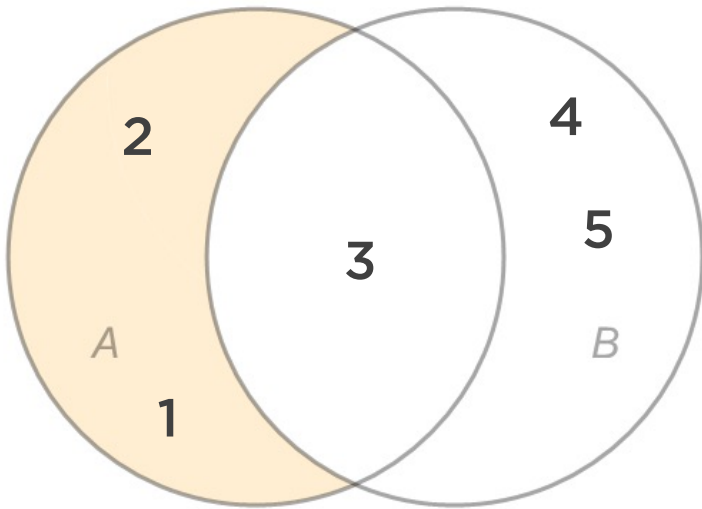
**Also exists as large operator**

$$\bigcap_{i=1}^3 A_i = A_1 \cap A_2 \cap A_3$$





# Set Difference



Given two sets  $A$  and  $B$ , the *set difference* is the set of items in  $A$  that are *not* in  $B$

For example, if  $A = \{1,2,3\}$  and  $B = \{3,4,5\}$  then

$$A \setminus B = \{1,2\}$$

Some obvious equalities

$$A \setminus A = \emptyset$$

$$\emptyset \setminus A = \emptyset$$

$$A \setminus \emptyset = A$$

# Cardinality and Set Complement

**Set *cardinality*** = number of elements in a set

**E.g.**, if  $A = \{a, b, c\}$ ,  $\#A = 3$

**Set *complement*** = set of elements not contained in  $A$  (but contained elsewhere)

**E.g.**, if  $E = \{2, 4, 6, \dots\}$  is a set of even natural numbers, then  $E^c = \{1, 3, 5, \dots\}$  is the set of odd natural numbers

**This can also be represented as**

$$E^c = \mathbb{N} \setminus E$$

**Die roll:** if  $A = \{1, 2, 5\}$ , then, assuming  $S = \{1, 2, 3, 4, 5, 6\}$ ,  $A^c = S \setminus A = \{3, 4, 6\}$



# Some Set Laws

## De Morgan's Laws

$$(A \cup B)^c = A^c \cap B^c$$

$$(A \cap B)^c = A^c \cup B^c$$

## Distributive Properties

$$A \cap (B \cup C) = (A \cap B) \cup (A \cap C)$$

$$A \cup (B \cap C) = (A \cup B) \cap (A \cup C)$$

**For every two sets  $A$  and  $B$ ,**

- $A \cap B$  and  $A \cap B^c$  are disjoint
- $A = (A \cap B) \cup (A \cap B^c)$

**A lot more laws and set operations; consult a textbook**



# Demo



## Set Theory in R

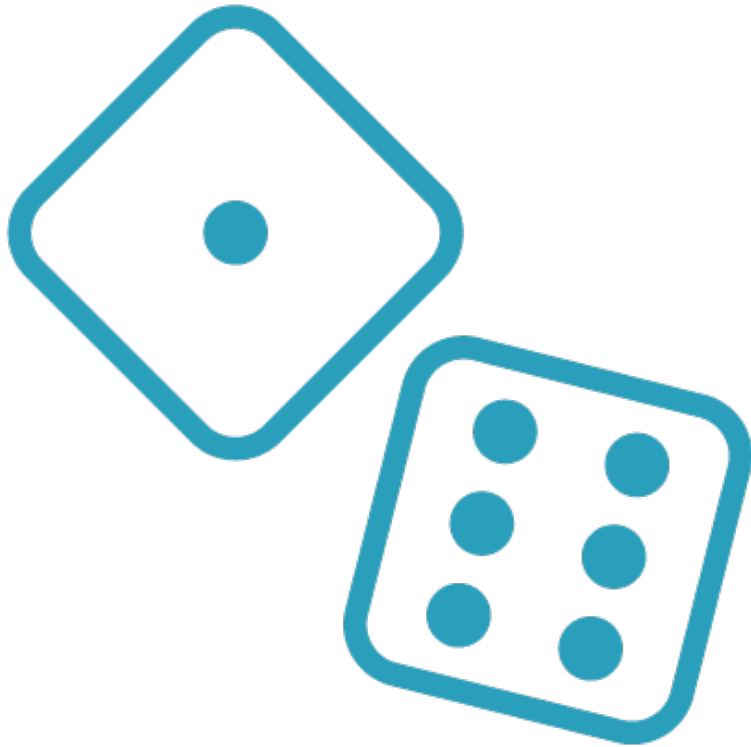


# Experiments and Events

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# Events



Consider an experiment consisting of a single balanced 6-sided die roll

The value rolled is a *random variable*

**Player rolled a 6**

- This is a *simple event*: it cannot be decomposed

**Player rolled an even number**

- Player rolled a 2, 4 or 6
- This is a *complex event*: it can be decomposed into simpler events

# Sample Space

The set of all possible values of a random variable is called the *sample space*.

The sample space for a die roll is

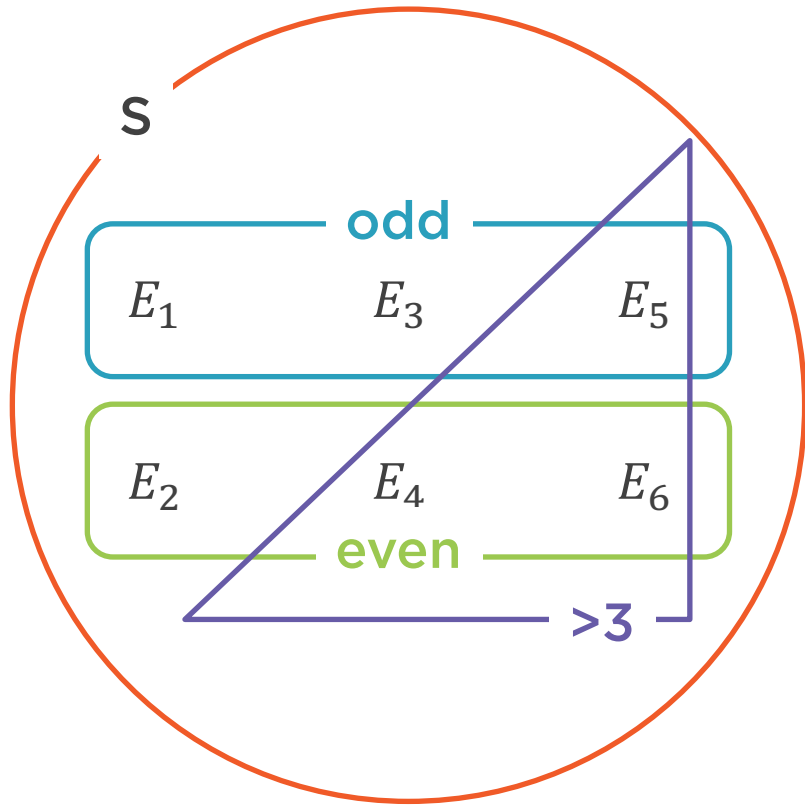
$$S = \{E_1, E_2, E_3, E_4, E_5, E_6\}$$

where  $E_n$  is an event when  $n$  is rolled.

Each element of the sample space is a *sample point*.



# Set Operations on Events



Suppose a die roll is

$$S = \{E_1, E_2, E_3, E_4, E_5, E_6\}$$

An even roll

$$S_{\text{even}} = \{E_2, E_4, E_6\}$$

An roll that is greater than 3

$$S_{>3} = \{E_4, E_5, E_6\}$$

A roll that is even or greater than 3

$$S_{\text{even}} \cup S_{>3} = \{E_2, E_4, E_5, E_6\}$$

A roll that is even and greater than 3

$$S_{\text{even}} \cap S_{>3} = \{E_4, E_6\}$$

A roll that is odd and greater than 5

$$S_{\text{odd}} \cap S_{>5} = \emptyset$$



# Independence



If events do not influence one another, they are called *independent*

- Successive die rolls or coin flips

If subsequent events influence one another, they are *dependent*

- Colored balls being pulled out of a hat without replacement
- The set of possible balls gets reduced as you pull them out

# Probability

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# Probability

How likely an event is to occur.



# Probability

**Probability of an event is a number that is**

- A value between 0 and 1 inclusive
- Value of 0 corresponds to “unlikely”, “almost never”
- Value of 1 corresponds to “almost surely”, abbreviated a.s.

**Typically recorded as  $P(\text{event}) = \text{value}$**

**Sometimes also expressed as percentage (0% to 100% respectively)**

- There's a 50% chance that  $\rightarrow P = 0.5 = \frac{1}{2}$
- I am 90% certain that  $\rightarrow P = 0.9 = \frac{9}{10}$
- Used in e.g., Excel



## Rules of Probability

Probability of event A is greater than or equal to zero

$$P(A) \geq 0$$

Probability of sample space is one

$$P(S) = 1$$

**If  $A_1, A_2, A_3, \dots$  are a sequence of mutually exclusive events ( $A_i \cap A_j = \emptyset$  if  $i \neq j$ ), then**

$$P(A_1 \cup A_2 \cup A_3 \cup \dots) = \bigcup A_i = \sum P(A_i)$$

## Probability Examples

Consider a coin flip  $S = \{E_H, E_T\}$

$P(S) = 1$  by definition

$P(E_H) = P(E_T)$  as both events are equally likely

Solving  $\begin{cases} P(E_H) + P(E_T) = 1 \\ P(E_H) - P(E_T) = 0 \end{cases}$  gives us

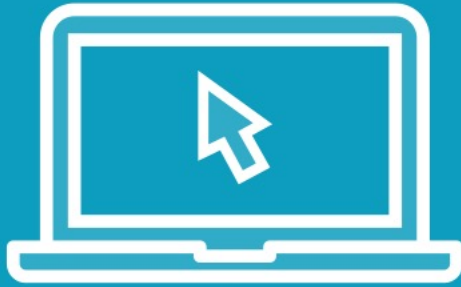
$$P(E_H) = P(E_T) = \frac{1}{2}$$

$P(\text{head and tail}) = 0$  (coin cannot on both head and tail at the same time)

$P(\text{head or tail}) = P(S) = 1$  a.s.  
(coin will definitely land on head or tail)



# Demo



## Basic Probability Simulation



# Discrete vs Continuous

## Discrete = finite set of unique values

- Coin toss, die roll, number of cars in a household
- $P(E) = \frac{1}{N}$  given  $N$  possible events, assuming all events equally likely

## Continuous = infinite set of values

- Person's height, amount of rainfall in a day
- $P(\text{you are 1.77m tall}) = 0$  unless you round people's height
- Instead, better to measure intervals, e.g.,  $P(1.77 < h < 1.78)$





# Counting Methods

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## Counting Sample Points

Consider a coin flip  $S = \{E_H, E_T\}$

Flip a coin three times (or flip 3 coins once)

- Sampling *with* replacement

How many different arrangements are possible?

Direct approach: list all the arrangements and count them

HHH, HHT, HTH, HTT, THH, THT, TTH, TTT

How many arrangements begin or end with a head?  $\#(S_{H??} \cup S_{??H})$

Again, count them: 6

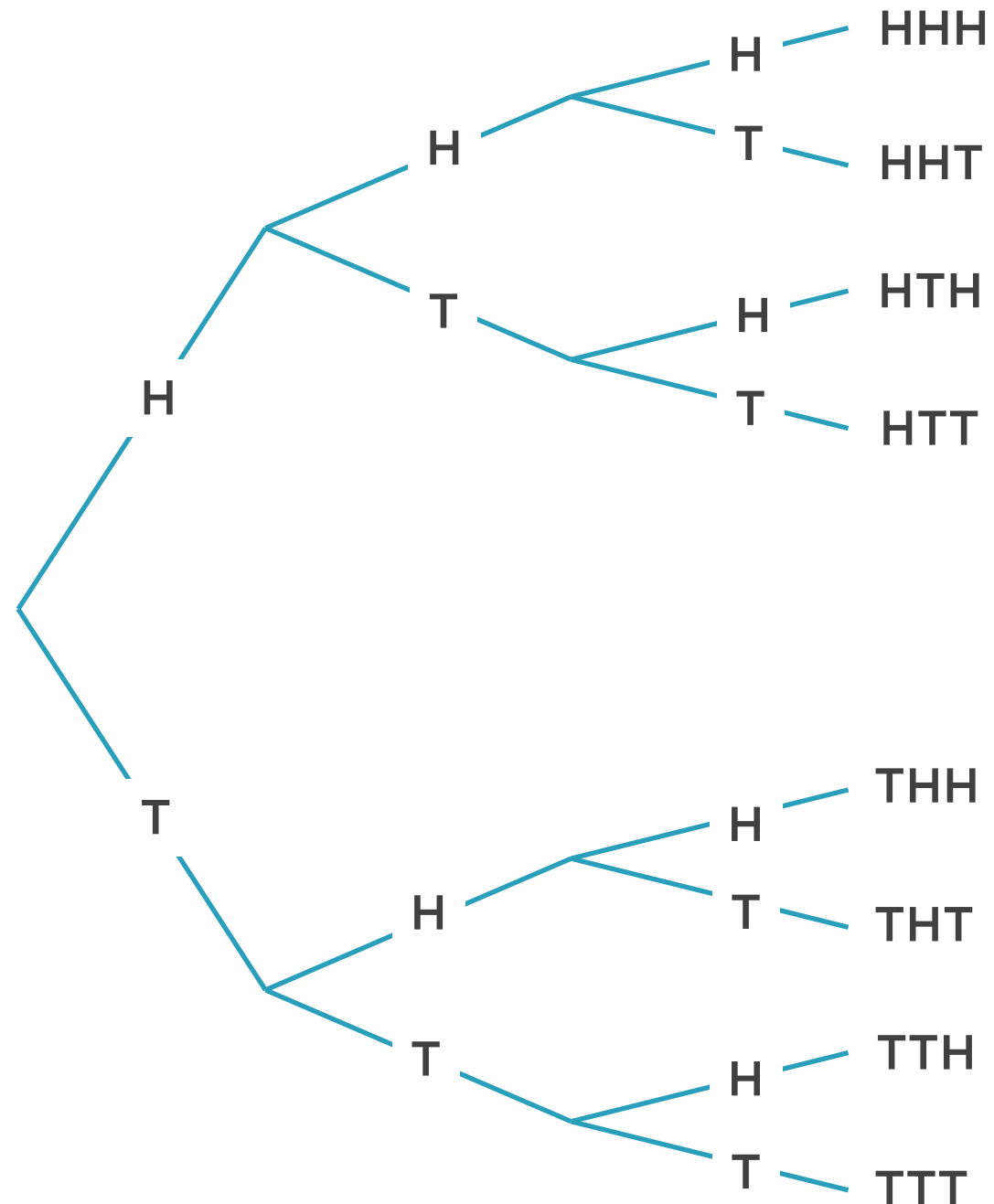


We can also visualize the sample space as a tree

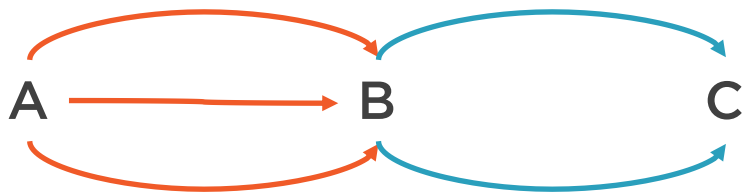
Each branch represents a possible event

Final branches represent all possible outcomes

Clearly, given  $N$  tosses, total # of outcomes =  $2^N$



# Multiplication Rule



We need to travel from A to C through B

Three ways of getting from A to B

$$- AB = \{AB_1, AB_2, AB_3\}$$

Two ways of getting from B to C

$$- BC = \{BC_1, BC_2\}$$

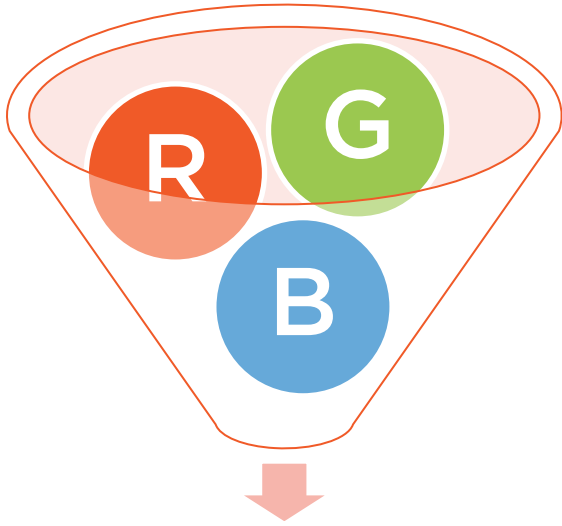
How many ways of getting from A to C?

	$AB_1$	$AB_2$	$AB_3$
$BC_1$	$(AB_1, BC_1)$	$(AB_2, BC_1)$	$(AB_3, BC_1)$
$BC_2$	$(AB_1, BC_2)$	$(AB_2, BC_2)$	$(AB_3, BC_2)$

Clearly, number of options is

$$\#AB \cdot \#BC = 3 \cdot 2 = 6$$





A bag has three balls: red, green, blue

You draw each of the balls in turn *without replacement* (you keep the ball)

How many possible draws are there?

First draw: three possible options

Second draw: one ball already taken, two possible options

Last draw: only one option

Thus, total number of possible draws is:

$$3 \times 2 \times 1 = 6$$

They are: RGB, RBG, GRB, GBR, BRG, BGR



# Permutations

Given a set of elements, all distinct arrangements of these elements are called the *permutations* of a set

The number of ways you can arrange N elements out of N possibilities is...

$$P_{n,n} = N \times (N - 1) \times N - 2 \times \dots \times 2 \times 1 = N!$$

**permutations**

**N! = “N factorial” = a product of all numbers from N down to 1**

- Assumes N is a non-negative integer
- By definition,  $0! = 1$



## General Formula for Permutations

Suppose there are  $n = 5$  balls (RGBWO) and you draw  $k = 3$  without replacement

How many possible arrangements are there?

5 possibilities on first draw, 4 on the second, 3 on the third, so  $5 \times 4 \times 3 = 60$

But how to express it in terms of  $n$  and  $k$ ?

$$5 \times 4 \times 3 = \frac{5 \times 4 \times 3 \times 2 \times 1}{2 \times 1} = \frac{5!}{2!} = \frac{n!}{(n-k)!}$$

Thus, the number of permutations of  $k$  elements out of  $n$  is  $P_{n,k} = \frac{n!}{(n-k)!}$



# Permutation Examples

Number of unique 5-letter words that can be arranged from letters  $\{a, b, c, d, e\}$

$$5^5 = 3125$$

If each letter appears exactly once

$$5! = 120$$

3-letter words, each letter appears once

$$P_{5,3} = \frac{5!}{(5-3)!} = 60$$





## Birthday Problem

Given a room with  $k$  people, what is the probability that *at least* two people have the same birthday?

- Ignore leap years, seasonal variations

Obvious: if  $k > 365$ ,  $P = 1$ , so assume  $k \leq 365$

Number of arrangements when each birthday is different is  $P_{365,k}$

Total number of arrangements =  $365^k$

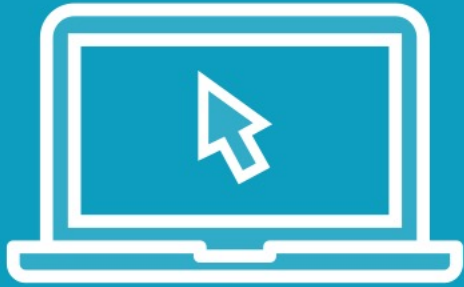
$$P(\text{no two people have same b/day}) = \frac{P_{365,k}}{365^k}$$

$\therefore P(\text{two+ people have same b/day})$

$$= 1 - \frac{P_{365,k}}{365^k}$$



# Demo



## Birthday Problem



# Combinatorics

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# Combinatorics

Combinatorics is a branch of mathematics concerning the study of finite or countable discrete structures.



## Combinatorial methods

Given a bag of balls  $B = \{R, G, B, W, O\}$

If we pull out 2 balls after replacement, the number of arrangements is

$$P_{5,2} = \frac{5!}{3!} = 20$$

What if we don't care about order, i.e., we consider  $\{R, B\}$  and  $\{B, R\}$  to be the same pick?

How can we calculate the number of unique picks now?

Problem: find the number of subsets of a set (remember, order doesn't matter!)



## Combinatorial methods

Given a bag of balls  $B = \{R, G, B, W, O\}$

The total number of picks (including duplicate sets) is  $P_{5,2}$

- And each of the picks has  $2!$  possible arrangements

So we reduce the total number of picks by the number of arrangements in each pick

$$C_{n,k} = \frac{P_{n,k}}{k!} = \frac{n!}{k! (n-k)!}$$

In our case,  $C_{5,2} = \frac{5!}{2!(5-2)!} = 10$

Here they are:

$\{RG, RB, RW, RO, GB, GW, GO, BW, BO, WO\}$



# Binomial Coefficients

We also denote  $C_{n,k}$  as  $\binom{n}{k}$

$\binom{n}{k}$  is called a *binomial coefficient* because, according to the Binomial Theorem,

$$(x + y)^n = \sum_{k=0}^n \binom{n}{k} x^k y^{n-k}$$

$$\forall n, \binom{n}{0} = \binom{n}{n} = 1$$

$$\text{Also, } \binom{n}{k} = \binom{n}{n-k}$$

## Binomial Coefficient Examples

You toss a coin 10 times and record the result. What is the probability of getting exactly 4 heads?

There are a total of  $2^{10}$  arrangements

Each arrangement is a choice as to where put the 4 heads among 10 tosses;  $\binom{10}{4}$  arrangements

$$\therefore P(4\text{heads}) = \frac{\binom{10}{4}}{2^{10}} = 0.205$$

Number of sequences of 4 heads *or less*

$$P(\leq 4\text{heads}) = \frac{\sum_{i=0}^4 \binom{10}{i}}{2^{10}} = 0.377$$





## Binomial Coefficient Examples

A class has 15 girls and 30 boys

Pick 10 children at random. What's the probability you'll pick exactly 3 girls?

Number of ways of picking 3 girls from 15 girls is  $\binom{15}{3}$ ; number of ways of picking 7 boys from 30 is  $\binom{30}{7}$

Overall number of combinations is  $\binom{45}{10}$

$$\therefore P(3 \text{ girls}) = \frac{\binom{15}{3}\binom{30}{7}}{\binom{45}{10}} = 0.29$$



## Multinomial Coefficients

10 students need to form 3 groups  
consisting of 4, 3 and 3 members  
respectively

How many ways can students be assigned  
to these groups?

First group: choose 4 students out of 10,  
 $\binom{10}{4}$  arrangements

We are left with 6 students; number of  
ways to split them is  $\binom{6}{3}$

$$\binom{10}{4}\binom{6}{3} = \frac{10!}{4!6!} \cdot \frac{6!}{3!3!} = \frac{10!}{4!3!3!} = 4200$$



# Multinomial Coefficients

In general, number of arrangements of  $n$  elements into  $k$  groups of size  $n_i = \{n_1, n_2, \dots\}$  is

$$\binom{n}{n_1} \binom{n - n_1}{n_2} \binom{n - n_1 - n_2}{n_3} \dots = \frac{n!}{n_1! n_2! \dots n_k!}$$

This is the *multinomial coefficient*, written as

$$\binom{n}{n_1, n_2, \dots, n_k}$$

Just like with binomial coefficients,

$$(x_1 + \dots + x_k)^n = \sum \binom{n}{n_1, n_2, \dots, n_k} x_1^{n_1} x_2^{n_2} \dots x_k^{n_k}$$



# Multinomial Coefficient Examples

**Picking 4,3,3 students out of 10**

$$\binom{10}{4,3,3} = \frac{10!}{4! 3! 3!} = 4200$$

**Number of ways to arrange 3 a's, 4 b's and 5 c's is**

$$\binom{12}{3,4,5} = \frac{12!}{3! 4! 5!} = 27,720$$

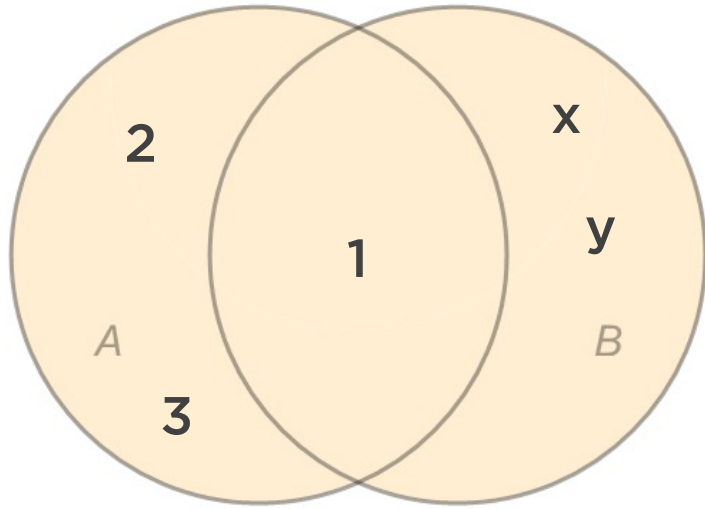
**In the expansion of  $(x + y + z)^3$ , the coefficient of the term  $x^2z = x^2y^0z^1$  is**

$$\binom{3}{2,0,1} = \frac{3!}{2! 0! 1!} = 3$$

# Probability of a Union of Events

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We also know that for a set of disjoint events  $A_i$

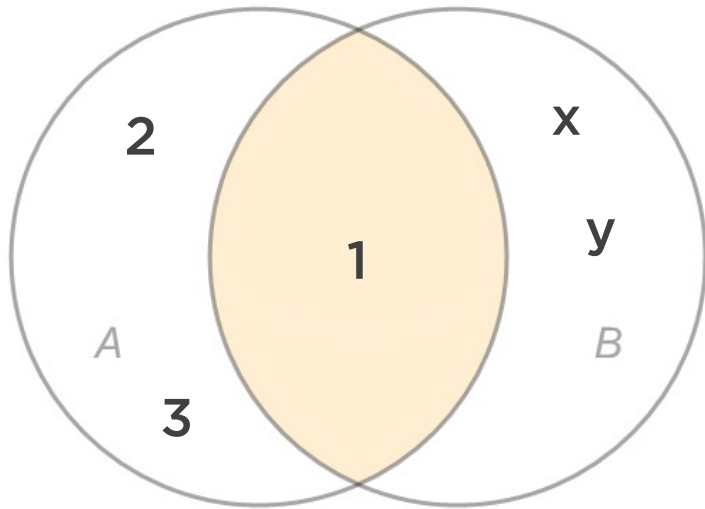
$$P\left(\bigcup A_i\right) = \sum P(A_i)$$

For every two (not necessarily disjoint) events  $A$  and  $B$ ,

$$P(A \cup B) = P(A) + P(B) - P(A \cap B)$$

This can be extended to 3 or more events, e.g.,

$$\begin{aligned} &P(A \cup B \cup C) \\ &= P(A) + P(B) + P(C) \\ &\quad - \{P(A \cap B) + P(B \cap C) + P(A \cap C)\} \\ &\quad + P(A \cap B \cap C) \end{aligned}$$



## Union of Events Example

Consider a cohort of 200 students

50 students take programming (P), 100 students take electronics(E), 150 students take maths (M)

30 students take programming+electronics, 45 students take electronics+maths, 25 students take electronics+programming

15 students take all 3 classes; some students take no classes from the list

What is the probability that a student takes at least one class?



## Union of Events Example

$$P(P) = \frac{50}{200}, P(E) = \frac{100}{200}, P(M) = \frac{75}{200}$$

$$P(P \cap E) = \frac{30}{200}, P(E \cap M) = \frac{45}{200}, P(E \cap P) = \frac{25}{200}$$

$$P(P \cap E \cap M) = \frac{15}{200}$$

$$\begin{aligned}\therefore P(P \cup E \cup M) &= \frac{50}{200} + \frac{100}{200} + \frac{75}{200} - \left\{ \frac{30}{200} + \frac{45}{200} + \frac{25}{200} \right\} + \frac{15}{200} \\ &= \frac{140}{200} = 0.7\end{aligned}$$



# Summary



To study phenomena, we perform experiments and record our observations

The sample space describes all possible outcomes; these can be modeled with sets and measured using counting or combinatorics methods

Probability: value in range 0 to 1 inclusive; described the likelihood of an event occurring

If events are mutually independent,  
 $P(\text{union of events}) = \text{sum of their individual probabilities}$

