

CSC380: Principles of Data Science

Probability 1

Xinchen Yu

- What is probability?
- Events
- Calculating probabilities
- Set Theory
- Law of Total Probability

What is probability?

Interpreting probabilities

Basically two different ways to interpret:

- Objective probability
 - based on logical analysis or long-run frequency of an event. It's derived from known facts, symmetry, or repeated experiments.
- Subjective probability
 - based on personal belief, opinion, or information about how likely an event is, especially when there's uncertainty or limited data.

Outcome, Event and Probability



Outcome

- Outcome is a single result/observation of a random experiment.
- Example 1: You flip a coin once.
 - The outcomes are: "Heads" or "Tails"
- Example 2: You roll a 6-sided die.
 - The outcomes are: 1, or 2, or 3, or 4, or 5, or 6
- Example 3: You tap "shuffle play" on your favorite Spotify playlist with 100 songs and the first song played.
 - The outcomes are: the 1st song in the list, or the 2nd song,

The Sample Space

- The set of all possible outcomes of a random experiment is called the sample space, written as S .
- In math, the standard notation for a set is to write the individual members in curly braces:
 - $S = \{\text{Outcome1}, \text{Outcome2}, \dots, \}$
- Useful to visualize the sample space with an actual space.

Events

- An event E is a **subset** of the sample space.
- An event E is a **set** of outcomes
- When we make a particular observation, it is either “in” E or not.
Helpful to think about events as propositions (TRUE/FALSE).
 - The proposition is TRUE when the outcome is among the elements of the event set, and FALSE otherwise.
 - Is 4 in event $E = \{2, 4, 6\}$?  YES \rightarrow the **proposition is TRUE**
 - Is 4 in event $F = \{1, 3, 5\}$?  NO \rightarrow the **proposition is FALSE**

Special events

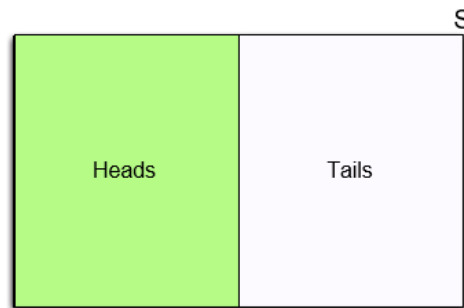
- The sample space (S) includes all possible outcomes.
 - If an event $E = S$, then no matter what outcome occurs it's always in E .
 - e.g., $E = \{\text{Heads, Tails}\}$
- The empty set \emptyset is also an event
 - It is an event that never happens
 - e.g. “the die comes up 7”, $E = \{7\}$

Calculating Probabilities

Calculating probability using symmetry

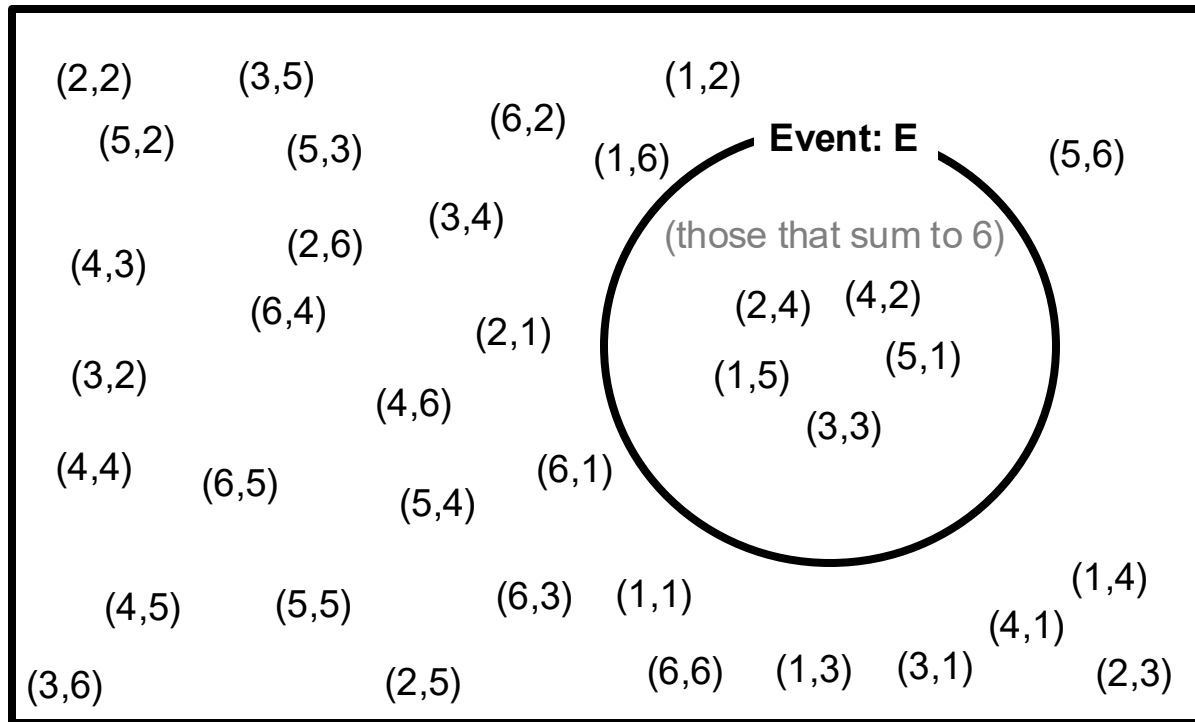
- If we have a sample space for which every outcome is equally likely, then we can find event probabilities easily.
- Since every outcome has the same “area”, we can just count:

$$P(E) = \frac{\text{\#outcomes in } E}{\text{\#outcomes in } S}$$



Random Events and Probability

What is the probability of having two numbers sum to 6?



$$P(E) = \frac{\text{\#outcomes in } E}{\text{\#outcomes in } S}$$

$$S = \{(a, b) : a, b \in \{1, \dots, 6\}\}$$

Each outcome is equally likely

of outcomes that sum to 6:

5

answer:

$$5/36 = 0.13888\dots$$

Probability as Area

- Notice that we can find the total probability of an event by breaking it into pieces and adding up the probabilities of the pieces:

$$P(\text{Even}) = P(2) + P(4) + P(6)$$

1	2 $P(2) = 1/6$	3
4 $P(4) = 1/6$	5	6 $P(6) = 1/6$

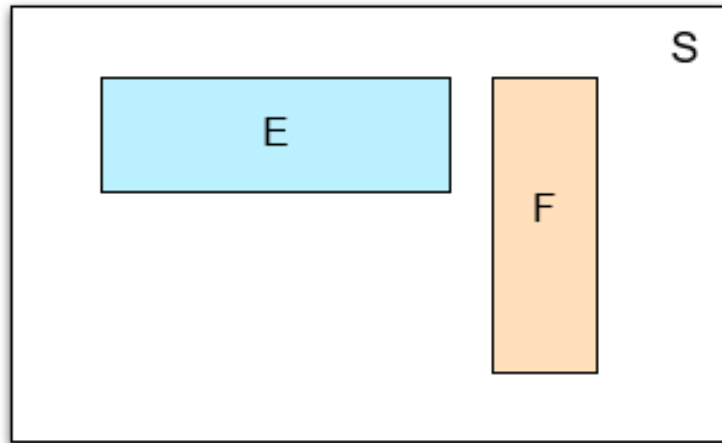
- These pieces are called 'elementary events'
 - Events that correspond to exactly one outcome

Probability as Area

- In general, breaking an event into *disjoint events* preserves the total probability
- **Disjoint:** E and F are *disjoint* if they cannot happen simultaneously,
- e.g. $E = \{2, 4\}$, $F = \{1, 3, 5\}$

- In such cases,

$$P(E \text{ or } F) = P(E) + P(F)$$



Partition

- We say that disjoint events E_1, \dots, E_n form a *partition* of E if any outcome in E lies in exactly one E_i
- e.g.
 - {Fr.} E_1 , {Soph.} E_2 form a partition of {Lower division} E
 - {Fr.}, {Soph.} {J.} {Sen.} form a partition of S (sample space)

Freshmen	Sophomores	Juniors	Seniors
----------	------------	---------	---------

Probability as Area

For disjoint events E, F :

$$P(E \text{ or } F) = P(E) + P(F)$$

More generally, If disjoint events E_1, \dots, E_n forms a partition of :

$$P(E) = P(E_1) + P(E_2) + \dots + P(E_n)$$

For event F (law of total probability):

$$P(F) = P(E_1, F) + P(E_2, F) + \dots + P(E_n, F)$$

Notation: $P(A, B)$ is a shorthand for $P(A \text{ and } B)$

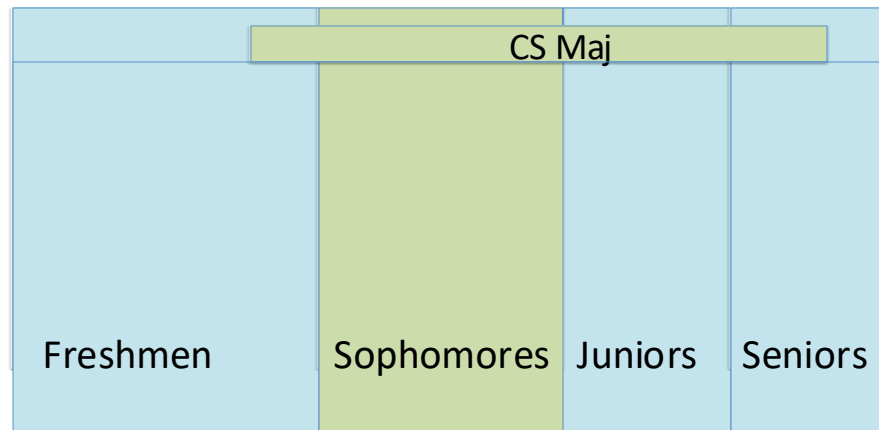
Inclusion-Exclusion Principle

Inclusion-Exclusion Principle For any events E and F ,

$$P(E \text{ or } F) = P(E) + P(F) - P(E \text{ and } F)$$

Accounting for overlap
between E and F

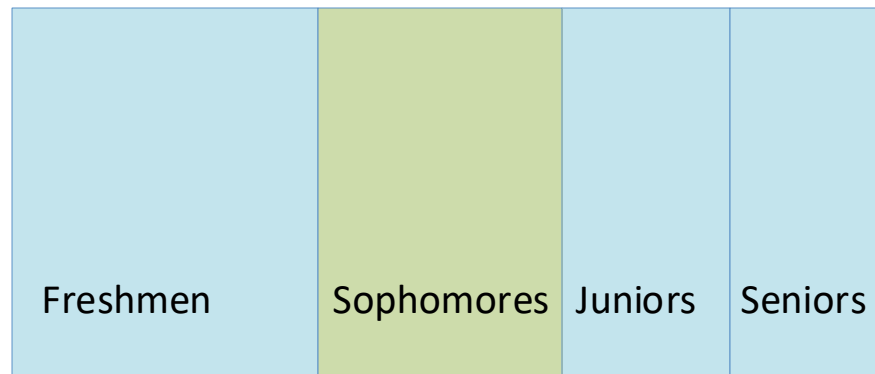
S



Complementary events

How would I find $P(\text{Non-Sophomore})$?

- Could just list the non-sophomores and then count, but we can use the fact that $P(S) = 1$ and *subtract* instead.
 - $P(\text{Non-Sophomore}) = 1 - P(\text{Sophomore})$



Set operations

Events and Set Theory

An event is a set of outcomes and a subset of sample space, so we use set theory to describe and combine them.

Two dice example:

E_1 : *First* die rolls 1

$$E_1 = \{(1, 1), (1, 2), \dots, (1, 6)\}$$

E_2 : *Second* die rolls 1

$$E_2 = \{(1, 1), (2, 1), \dots, (6, 1)\}$$

- Any die rolls 1: event E_1 or E_2
- Set operation: $E_1 \cup E_2$

Set operations



Two dice example:

E_1 : First die rolls 1

E_2 : Second die rolls 1

$$E_1 = \{(1, 1), (1, 2), \dots, (1, 6)\}$$

$$E_2 = \{(1, 1), (2, 1), \dots, (6, 1)\}$$

Operators on events:

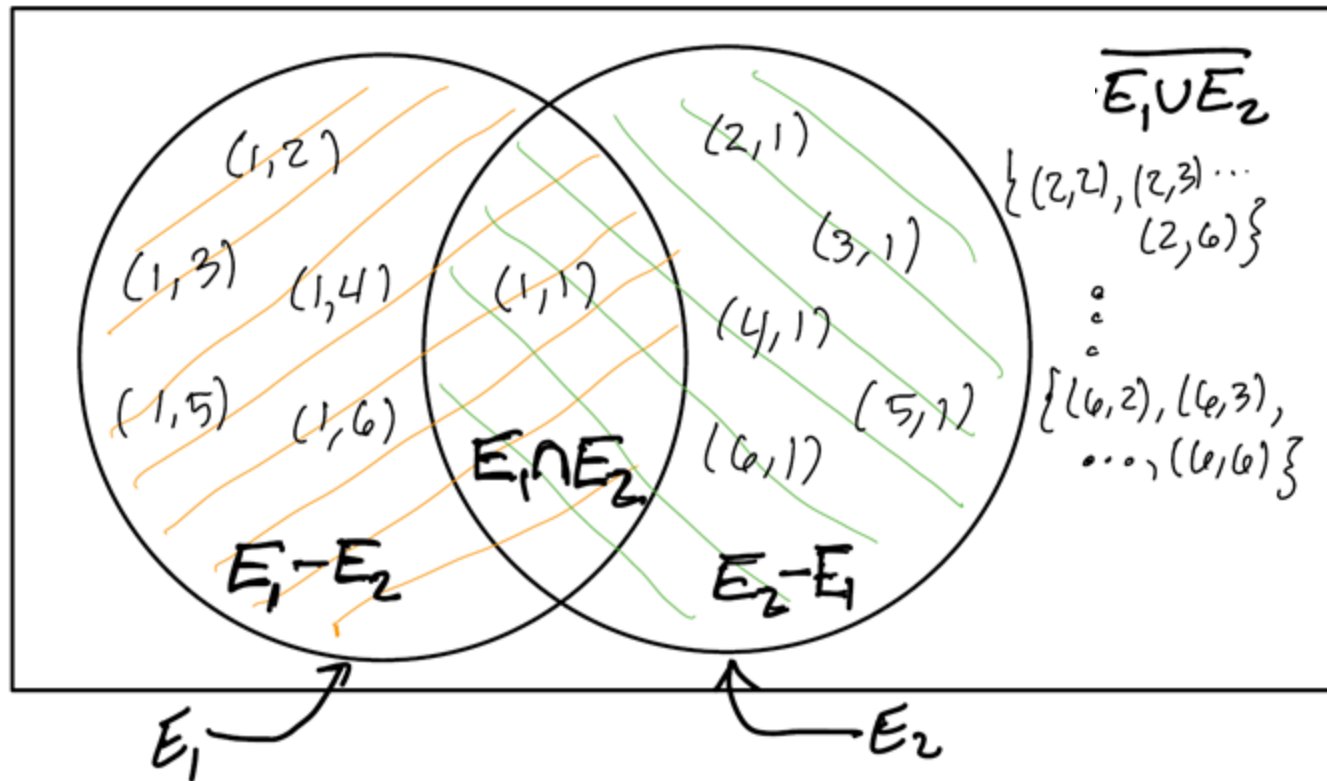
Operation	Value	Interpretation
$E_1 \cup E_2$	$\{(1, 1), (1, 2), \dots, (1, 6), (2, 1), \dots, (6, 1)\}$	Any die rolls 1
$E_1 \cap E_2$	$\{(1, 1)\}$	Both dice roll 1
$E_1 \setminus E_2$	$\{(1, 2), (1, 3), (1, 4), (1, 5), (1, 6)\}$	Only the first die rolls 1
$\overline{E_1 \cup E_2}$	$\{(2, 2), (2, 3), \dots, (2, 6), (3, 2), \dots, (6, 6)\}$	No die rolls 1

$$(\text{= } E_1 - E_2 \text{ := } E_1 \cap E_2^c)$$

$$(\text{= } (E_1 \cup E_2)^c)$$

Set operations

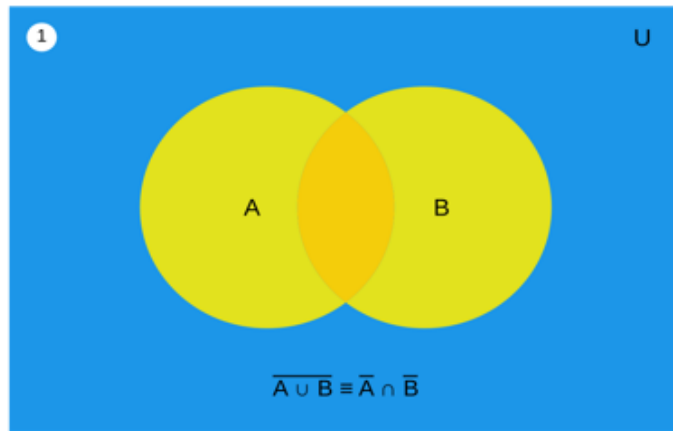
Can interpret these operations using a Venn diagram...



De Morgan Law 1 $(A \cup B)^C = A^C \cap B^C$

Example:

- A: I bring my cellphone
- B: I bring my laptop
- A^C : I don't bring my cellphone
- B^C : I don't bring my laptop

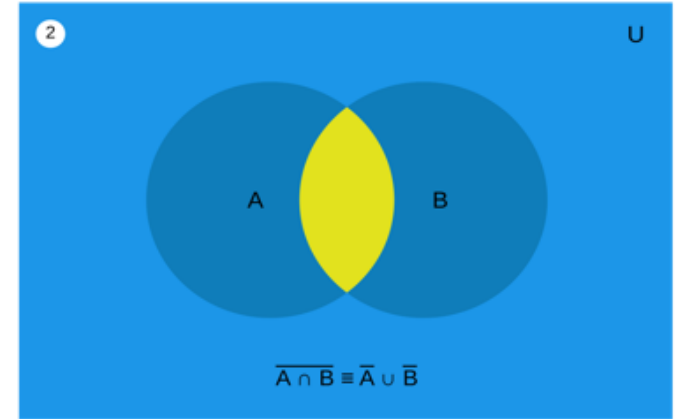


- $A \cup B$: I bring my cellphone or my laptop
- $(A \cup B)^C$: I bring neither my cellphone nor my laptop
- $A^C \cap B^C$: I didn't bring my cellphone & I didn't bring my laptop

Set Theory: De Morgan Law

De Morgan Law 2 $(A \cap B)^c = A^c \cup B^c$

Ex: try to make sense of it using the same example above

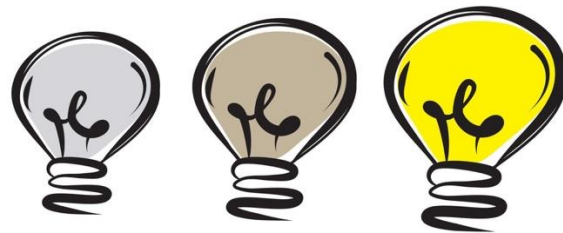


De Morgan Law generalizes to a collection of n events

- But first, let's define some notations

Intersection / union over n events

- n lightbulbs
- E_i : i -th lightbulb is on



- How to describe the event that at least one lightbulb is on?
 - i.e. bulb 1 is on OR ... OR bulb n is on

$$E_1 \cup \cdots \cup E_n =: \bigcup_{i=1}^n E_i$$

- How to describe the event that all lightbulbs are on?

$$E_1 \cap \cdots \cap E_n = \bigcap_{i=1}^n E_i$$

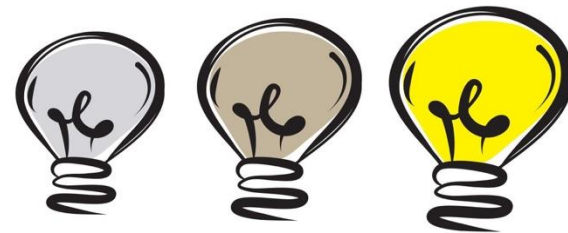
De Morgan Laws with n events

- De Morgan Laws:**

$$(E_1 \cup \dots \cup E_n)^c = E_1^c \cap \dots \cap E_n^c$$

Not (at least one bulb is on)

All bulbs are off



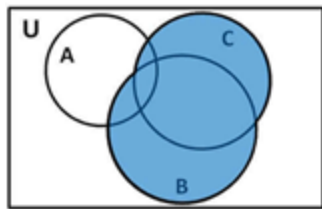
$$(E_1 \cap \dots \cap E_n)^c = E_1^c \cup \dots \cup E_n^c$$

Not (all bulbs are on)

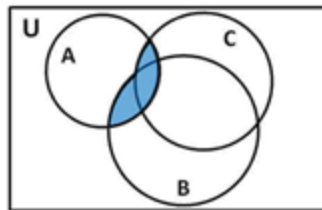
At least one bulbs is off

Set operation: Distributive law

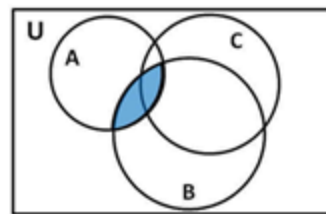
- Distributive law in arithmetics $a(x + y) = ax + ay$ carry over to sets
- **Distributive Law 1** $A \cap (B \cup C) = (A \cap B) \cup (A \cap C)$



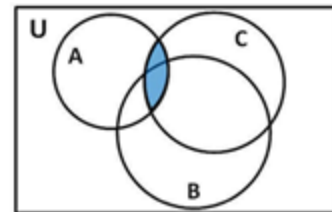
$(B \cup C)$



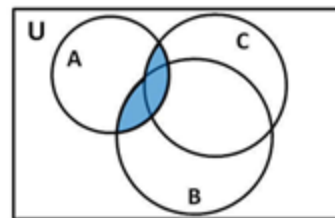
$A \cap (B \cup C)$



$(A \cap B)$



$(A \cap C)$



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

Set operation: Distributive law

- **Distributive Law 2** $A \cup (B \cap C) = (A \cup B) \cap (A \cup C)$
- Can justify this by:
 - drawing a picture (like previous slide), or
 - proving it using Distributive Law 1 and De Morgan Law

Rules of Probability

Rules of probability

- To recap and summarize:

Rules of Probability

- 1. Non-negativity:** All probabilities are between 0 and 1 (inclusive)
- 2. Unity of the sample space:** $P(S) = 1$
- 3. Complement Rule:** $P(E^C) = 1 - P(E)$
- 4. Probability of Unions:**
 - (a) In general, $P(E \cup F) = P(E) + P(F) - P(E \cap F)$*
 - (b) If E and F are disjoint, then $P(E \cup F) = P(E) + P(F)$*

Special case

Assume each outcome is **equally likely**, and sample space is finite, then the probability of event is:

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

Number of elements in event set

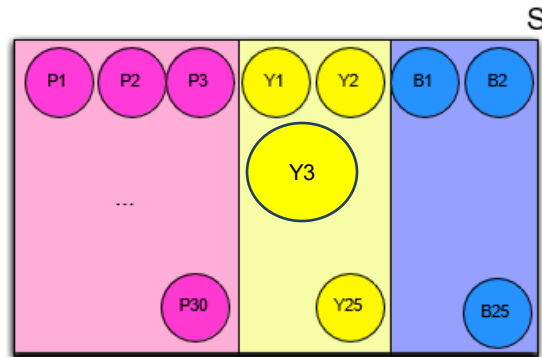
Number of possible outcomes (e.g. 36)



This is called classical probability model

Rethinking the classical probability model

- Classical probability model assumes all outcomes equally likely
- When is this applicable?
 - *Fair* coin toss, *fair* dice throw, ...
 - $S = \{P1, P2, \dots, P30, Y1, \dots, Y25, B1, \dots, B25\}$
- When is this assumption problematic?
 - *Unfair* coin toss (one side is heavier)
 - A yellow ball is much larger

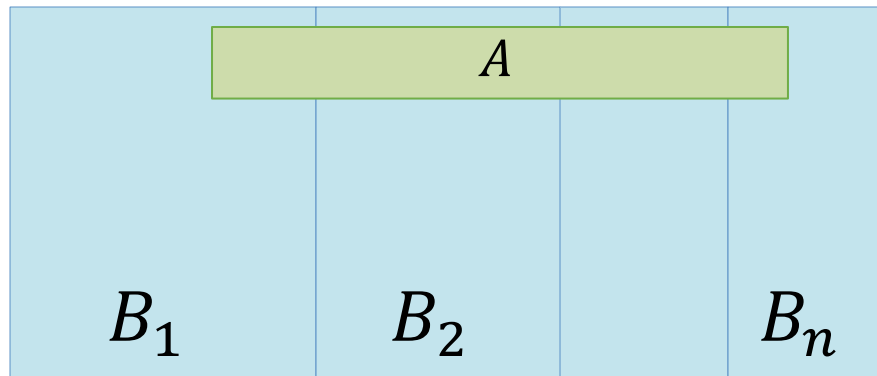


Law of Total Probability

Law of Total Probability

Law of Total Probability Suppose B_1, \dots, B_n form a partition of the sample space S . Then,

$$P(A) = P(A, B_1) + \dots + P(A, B_n)$$



- Recall notation: $P(A, B_1)$ is a shorthand for $P(A \cap B_1)$
- Why? $A \cap B_1, \dots, A \cap B_n$ form a partition of A

- If we know that all outcomes are equally likely, we can use

We will use combinatorics
to do counting

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

Number of elements
in event set

Number of possible
outcomes (e.g. 36)

- If $|E|$ is hard to calculate directly, we can try
 - the rules of probability
 - the Law of Total Probability, using an appropriate partition of sample space S