

STOR 565 Machine Learning  
Probability Overview

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# Basics of Set Theory

**Definition (informal):** A set is an unordered collection of objects.

**Notation:**  $A, B, C$  denote sets and  $x, y, z$  denote objects.

## Basic relations between sets and objects

### 1. Containment

- ▶ Notation  $x \in A$
- ▶ Meaning: Object  $x$  is a member/element of the set  $A$

# Basics of Set Theory, cont.

## 2. Subset

- ▶ Notation  $A \subseteq B$
- ▶ Set  $A$  is a subset of  $B$ . Formally, if  $x \in A$  then  $x \in B$ .

## 3. Equality

- ▶ Notation  $A = B$
- ▶ Sets  $A$  and  $B$  are equal if they have the same members.

**Note:**  $A = B$  is logically equivalent to  $A \subseteq B$  and  $B \subseteq A$ .

**Definition:** The **Cardinality** of a set  $A$ , written  $|A|$ , is the number of objects in  $A$ , which may be infinite.

## Sets Specified by *Lists*

**Template:** If  $x_1, x_2, \dots$  is a finite or infinite family of objects, then

$$\{x_1, x_2, \dots\}$$

is the set containing these objects, without regards to order or repetition.

### Examples:

- ▶  $A = \{1, 2, 5, 9\}$ ,  $|A| = 4$
- ▶  $B = \{\text{Ziggy, JoJo, Buddy}\}$ ,  $|B| = 3$
- ▶  $C = \{\text{All freshman at UNC}\}$ ,  $|C|$  is finite
- ▶ Natural numbers  $\mathbb{N} = \{0, 1, 2, 3, \dots\}$ ,  $|\mathbb{N}|$  is (countably) infinite
- ▶ Real numbers  $\mathbb{R} = (-\infty, +\infty)$ ,  $|\mathbb{R}|$  is (uncountably) infinite

**Note:**  $A = \{5, 2, 1, 9\}$  and  $B = \{\text{JoJo, Ziggy, Buddy, Ziggy}\}$

## The Empty Set and Recursive Bracing

**Definition:** The **empty set**, denoted  $\emptyset$ , is the set with no elements.

**Fact:** For any set  $A$ , it is vacuously true that  $\emptyset \subseteq A$ .

**Note:** If  $B$  is a set then  $\{B\}$  is the set that contains  $A$  as its sole element. Applying this rule recursively to the empty set  $\emptyset$ , we can define

$$A = \{\emptyset, \{\emptyset\}, \{\{\emptyset\}\}\}$$

### Questions:

- ▶ What is  $|A|$ ?
- ▶ Is  $\emptyset \subseteq A$ ? Is  $\emptyset \in A$ ?
- ▶ Is  $\{\emptyset\} \subseteq A$ ? Is  $\{\emptyset\} \in A$ ?

## Sets Specified by *Conditions*

**Template:** Wish to specify a subset of some universal/background set that satisfy some conditions. General format

$$\{x : \text{conditions}(x)\} = \text{The set of } x \text{ satisfying the conditions}$$

### Examples

- ▶  $\{x : x \text{ is prime}\} \subseteq \mathbb{N}$
- ▶ Rational numbers  $\{x : x = a/b \text{ where } a, b \in \mathbb{N} \text{ and } b > 0\} \subseteq \mathbb{R}$
- ▶ Unit interval  $[0, 1] = \{x : 0 \leq x \leq 1\} \subseteq \mathbb{R}$

## Basic Operations on Sets

**1. Complement**  $A^c = \{x : x \notin A\} =$  Objects *not* in  $A$

**2. Union**  $A \cup B = \{x : x \in A \text{ or } x \in B\} =$  Objects in  $A$  or  $B$

**3. Intersection**  $A \cap B = \{x : x \in A \text{ and } x \in B\} =$  Objects in  $A$  and  $B$

**Note:** Unions and intersections of three or more sets can be defined similarly

▶  $A \cup B \cup C$

▶  $A \cap B \cap C$ .

**Definition:** Sets  $A$  and  $B$  are **disjoint** if  $A \cap B = \emptyset$ , that is, they have no elements in common.

## Set Operations, cont.

**Basic containment relations:**  $A \cap B \subseteq A, B \subseteq A \cup B$

### DeMorgan's Laws

▶  $(A \cup B)^c = A^c \cap B^c$

▶  $(A \cap B)^c = A^c \cup B^c$

### Distributive Laws

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

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



# Random Experiments

An **experiment** is an idealized procedure that can be repeated and that has a well-defined set of possible outcomes.

An experiment is **random** if its outcome is uncertain in advance.

## Examples

- ▶ Flip a coin three times
- ▶ Roll a die two times
- ▶ Choose three cards at random from a standard deck
- ▶ Count number of defective items in shipment of electronic components
- ▶ Measure the lifetime of a particular brand of lightbulb

# Probability Model

Study of probability begins with a basic model for random experiments. The first two components of the model are:

1. **Sample space** Denoted  $\Omega$ , this is just the set of possible outcomes
2. **Events** An event is a (sometimes any) subset of the sample space

Terminology: An event  $A \subseteq \Omega$  **occurs** if outcome of the experiment is in  $A$ .

## Ex: Three Flips

**1. Sample space:** Let H = Heads, T = Tails. Then

$$\Omega = \{TTT, TTH, THT, HTT, THH, HTH, HHT, HHH\}$$

Note that  $|\Omega| = 2 \times 2 \times 2 = 8$ .

### 2. Some events

- ▶ No Heads =  $\{TTT\}$
- ▶ Number of Heads equals Number of Tails =  $\emptyset$
- ▶ First flip is Heads =  $\{HHH, HTH, HHT, HTT\}$
- ▶ At least two Heads =  $\{HHT, HTH, THH, HHH\}$
- ▶ All flips the same =  $\{HHH, TTT\}$

## Ex: Two Rolls

**1. Sample space:** Each roll shows 1, 2, 3, 4, 5, or 6 dots. Then

$$\Omega = \{(i, j) : 1 \leq i, j \leq 6\}$$

Here  $i$  = outcome of first roll,  $j$  = outcome of second roll.  $|\Omega| = 6 \times 6 = 36$ .

### Some events

- ▶ First roll is 1 =  $\{(1, 1), (1, 2), (1, 3), (1, 4), (1, 5), (1, 6)\}$
- ▶ Both rolls the same =  $\{(1, 1), (2, 2), (3, 3), (4, 4), (5, 5), (6, 6)\}$
- ▶ Sum of rolls is 9 =  $\{(3, 6), (6, 3), (4, 5), (5, 4)\}$
- ▶ Product of rolls is 8 =  $\{(2, 4), (4, 2)\}$

## Final Component of Probability Model

**Probability measure**  $P(\cdot)$ : For each event  $A \subseteq \Omega$

$P(A)$  = probability that the outcome of the experiment lies in  $A$

Probability measures satisfy the following **axioms**

1.  $P(A) \geq 0$  for every event  $A \subseteq \Omega$  (non-negative)
2.  $P(\Omega) = 1$  (sample space has probability one)
3. If  $A \cap B = \emptyset$  then  $P(A \cup B) = P(A) + P(B)$  (additivity)

**Basic Idea:** If we think of  $\Omega$  as a figure in the plane with unit area, then  $P(A)$  is like the area of  $A$ .

## Elementary Properties of Probabilities

1. If  $A \subseteq B$  then  $P(A) \leq P(B)$ .
2. For every event  $A$ , the probability  $P(A)$  is between 0 and 1.
3.  $P(A^c) = 1 - P(A)$
4.  $P(A \cup B) = P(A) + P(B) - P(A \cap B) \leq P(A) + P(B)$
5.  $P(\cup_{i=1}^n A_i) \leq \sum_{i=1}^n P(A_i)$

# Meaning/Interpretations of Probability

## 1. **Objective/Frequentist:** Long run relative frequency

If we repeat the basic experiment many times then for any event  $A$

$$\frac{\text{\# times } A \text{ occurs}}{\text{total \# repetitions}} \approx P(A)$$

Problem: How to interpret probability of singular events, e.g.,

- ▶ The Buffalo Bills win next year's Superbowl
- ▶ The next president of the U.S. is a woman

## 2. **Subjective/Bayesian:** Degree of belief or evidentiary support

Important: Mathematical truth of theorems of probability theory does not depend on interpretation.

## Uniform Probability Model

**Consider:** experiment with finite sample space  $\Omega = \{\omega_1, \dots, \omega_m\}$

**Uniform probability model:** All outcomes are equally likely, i.e.,

$$P(\omega_j) = \frac{1}{|\Omega|} \text{ for } 1 \leq j \leq m$$

In this case, the axioms ensure that for each event  $A \subseteq \Omega$

$$P(A) = \frac{|A|}{|\Omega|} = \frac{\# \text{ outcomes in } A}{\text{total \# outcomes}}$$

so we can find probabilities by counting outcomes.



## Ex: Three Flips

**Recall**  $\Omega = \{TTT, TTH, THT, HTT, THH, HTH, HHT, HHH\}$  and  $|\Omega| = 8$

### Consider events

$A = \text{First flip is Heads} = \{HHH, HTH, HHT, HTT\}$

$B = \text{Second flip is Heads} = \{HHH, THH, HHT, THT\}$

$A \cap B = \text{First two flips are Heads} = \{HHH, HHT\}$

Under the uniform probability model  $P(\cdot)$

$$P(A) = \frac{|A|}{|\Omega|} = \frac{4}{8} = \frac{1}{2} \qquad P(B) = \frac{|B|}{|\Omega|} = \frac{4}{8} = \frac{1}{2}$$

$$P(A \cap B) = \frac{|A \cap B|}{|\Omega|} = \frac{2}{8} = \frac{1}{4}$$

# Independence

**Definition:** Events  $A$  and  $B$  in a random experiment are **independent** if

$$P(A \cap B) = P(A) P(B)$$

**Note:**

- ▶ Independence is different from disjointness
  - ▶  $A$  and  $B$  are disjoint if they *don't* overlap
  - ▶  $A$  and  $B$  are independent if they overlap *just the right amount*
- ▶ Independence *depends on the probability*  $P(\cdot)$ .

## Ex: Three Flips

$A = \text{First flip is Heads} = \{\text{HHH, HTH, HHT, HTT}\}$

$B = \text{Second flip is Heads} = \{\text{HHH, THH, HHT, THT}\}$

$C = \text{At least two Heads} = \{\text{HHT, HTH, THH, HHH}\}$

Under uniform probability  $P()$ ,

$$P(A) \cdot P(B) = \frac{1}{2} \cdot \frac{1}{2} = \frac{1}{4} = P(A \cap B)$$

*Thus A and B are independent. By contrast,*

$$P(A \cap C) = \frac{|A \cap C|}{8} = \frac{3}{8} \neq P(A) \cdot P(C) = \frac{1}{4}$$

*Thus A and C are not independent.*

## Ex: Three Flips, continue

Let events  $A, B, C$  be as before, and consider the probability  $P'$  such that

$$P'(HHH) = P'(TTT) = 1/2$$

and all other outcomes have probability zero.

The probability  $P'()$  models an experiment in which the outcomes are **completely dependent**. Either all flips are  $H$  or all flips are  $T$ .

Under  $P'()$  the events  $A$  and  $C$  are not independent. In fact,

$$P(A) = P(B) = P(A \cap B) = 1/2$$

## Conditional Probability

**Definition:** Let  $A$  and  $B$  be events with  $P(B) > 0$ . The **conditional probability** of  $A$  given  $B$  is defined by

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

### Notes:

- ▶ The vertical bar '|' is read as "given"
- ▶  $B$  is the *conditioning event*; its probability goes in the denominator
- ▶ Interpretation of  $P(A | B)$ 
  - ▶ What is the probability of  $A$  if we are told that  $B$  occurred?
  - ▶ The proportion of  $B$  occupied by  $A$
  - ▶ The probability of  $A$  if  $\Omega$  were replaced by  $B$

## Conditional Probability, cont.

**Note:** If  $A$  and  $B$  are independent, then

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

and similarly,  $P(B|A) = P(B)$ .

**Interpretation:** Knowing that  $B$  occurred tells us nothing about the probability of  $A$ , and vice-versa.

In general,  $P(A|B)$  can be less than, equal to, or greater than  $P(A)$ .

## Example: Three Flips

### Consider events

$A$  = First flip is H

$B$  = Second flip is T

$C$  = At least two Hs

**Note:** Under uniform probability  $P$ , events  $A, B$  are independent, so

$$P(A | B) = P(A)$$

But events  $A, C$  and events  $B, C$  are **not** independent. In fact,

$$P(C | A) > P(C) \quad P(C | B) < P(C)$$

## Independence in Inference

**Problem:** Estimate (unknown) number  $N$  of people killed in a conflict

**Available data:** Two lists of individuals killed in the conflict, prepared by independent observers.

- ▶  $N_1$  = number of individuals on first list
- ▶  $N_2$  = number of individuals on second list
- ▶  $N_{1,2}$  = number of individuals on both lists

Consider **proportions**

$$\hat{p}_1 = \frac{N_1}{N} = \text{proportion of casualties on first list}$$

$$\hat{p}_2 = \frac{N_2}{N} = \text{proportion of casualties on second list}$$

$$\hat{p}_{1,2} = \frac{N_{1,2}}{N} = \text{proportion of casualties on both lists}$$



## Independence in Inference, cont.

### Key Ideas

- ▶ Proportions  $\hat{p}$  are estimates of true probabilities  $p$
- ▶ If the lists are collected independently, then  $p_{1,2} = p_1 \cdot p_2$ .

Substituting estimates  $\hat{p}$  for true probabilities  $p$  we find

$$\frac{N_{1,2}}{N} = \hat{p}_{1,2} = \hat{p}_1 \cdot \hat{p}_2 = \frac{N_1}{N} \cdot \frac{N_2}{N}$$

Rearranging terms gives estimate of number of fatalities

$$N \approx \frac{N_1 N_2}{N_{1,2}}$$

## Conditional Probability, cont.

**Chain Rule:** Rewriting definition of conditional probability gives

$$P(A \cap B) = P(A) \cdot P(B | A)$$

**Law of Total Probability:** Note that  $B = (B \cap A) \cup (B \cap A^c)$ . Then we have

$$\begin{aligned} P(B) &= P((B \cap A) \cup (B \cap A^c)) \quad \text{[by the identity above]} \\ &= P(B \cap A) + P(B \cap A^c) \quad \text{[as } B \cap A \text{ and } B \cap A^c \text{ are disjoint]} \\ &= P(B | A)P(A) + P(B | A^c)P(A^c) \quad \text{[by the chain rule]} \end{aligned}$$

## Bayes's Formula

**Fact:** If  $A$  and  $B$  are events with  $P(A), P(B) > 0$  then

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

**Proof:** By definition

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

Apply chain rule to the numerator, law of total probability to the denominator.

### Interpretation

- ▶  $P(A)$  is the **prior** probability of  $A$
- ▶  $P(A|B)$  is the **posterior** probability of  $A$  *after* we know  $B$  occurred
- ▶ Think of  $A$  as unobserved “cause” and  $B$  as observed “effect”
- ▶ Formula tells us how to update prior probability of  $A$  given  $B$

## Example: Guessing on Multiple Choice Question

Multiple choice question with 5 possible answers.

If a student knows the correct answer, they choose it; if not, they choose one of the answers at random (with equal probability).

Suppose that a student knows the correct answer with probability .70. If they answer the question correctly, how likely is it that they were guessing?

Cause (unobserved)  $A$  = student knows the correct answer

Effect (observed)  $B$  = student answers question correctly

## Multiple Choice Question, cont.

We are given

- ▶  $P(A) = .70$  so  $P(A^c) = 1 - P(A) = .30$
- ▶  $P(B | A) = 1$  and  $P(B | A^c) = 1/5 = .20$

Applying Bayes formula gives

$$\begin{aligned}P(A | B) &= \frac{P(B | A) P(A)}{P(B | A) P(A) + P(B | A^c) P(A^c)} \\ &= \frac{1 \times .70}{1 \times .70 + .20 \times .30} \\ &\approx .92\end{aligned}$$

## Boys and Girls

Family with two children: possible sexes of first, second children are

$$\{(M, M), (M, F), (F, M), (F, F)\}$$

Define events

$A$  = First child a girl

$B$  = Second child a girl

$C$  = At least one girl

Suppose all four possibilities are equally likely. Find

$$P(B | A)$$

$$P(B | C)$$

Note that  $P(C | B) = 1$  so  $P(C | B) \neq P(B | C)$ . Order of events matters!