CS145: Probability & Computing Lecture 1: Sets & Events, Counting & Combinatorics



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Figure credits: Bertsekas & Tsitsiklis, **Introduction to Probability**, 2008 Pitman, **Probability**, 1999

Suppose there are *m* students in a class. What is the probability that at least two students in the class have the same birthday?

In a class of 70 students, this probability is about 99.87%



CS145: Lecture 1 Outline

- Sample spaces: Sets of possible outcomes
- Probability: Counting and the Discrete Uniform Law
- Example: The birthday paradox



Defining a Probabilistic Space/Model



Elements of a Probabilistic Model

- The sample space Ω , which is the set of all possible outcomes of an experiment.
- The **probability law**, which assigns to a set A of possible outcomes (also called an **event**) a nonnegative number $\mathbf{P}(A)$ (called the **probability** of A) that encodes our knowledge or belief about the collective "likelihood" of the elements of A. The probability law must satisfy certain properties to be introduced shortly.

Background: Sets

- A set is a collection of objects, which are elements of the set.
- A set can be finite, $S = \{1, 2, \dots, n\}$. Cardinality (size): |S| = n
- A set can be countably infinite:

$$S = \{x \mid x = 2k + 1 \text{ or } x = -2k + 1, k \text{ integer} \}$$

= $\{1, -1, 3, -3, 5, -5, \dots\}.$

- A set can be uncountable, $S = \{x \mid x \in [0, 1]\}$.
- A set can be empty $S = \emptyset$.

Sets: Elements & Relationships

- $x \in S$ the element x is a member of the set S
- $x \notin S$ the element x is not a member of the set S
- $\exists x$ there exists x...
- \forall for all elements \mathbf{x} ...
- $T \subseteq S$ $\forall x \in T$, $x \in S$
- $T \subset S \forall x \in T$, $x \in S$ AND $\exists x \in S$ such that $x \notin T$.

Sets: Combination & Manipulation

- A base set Ω , all sets are subsets of Ω
- Basic operations: for $S, T \subseteq \Omega$,
 - $S \cup T = \{x \mid x \in S \text{ or } x \in T\}$
 - $\underline{S} \cap T = \{x \mid x \in S \text{ and } x \in T\}$
 - $\overline{S} = S^c = \{x \mid x \notin S\}$
- De Morgan's laws:
 - $(S \cup T)^c = \overline{S} \cap \overline{T}$ • $(S \cap T)^c = \overline{S} \cup \overline{T}$





Visualizing Sets: Venn Diagrams



Partitions of a Set



A set B is *partitioned* into *n* subsets if:

$$B_1 \cup B_2 \cup \dots \cup B_n = B$$
$$B_i \cap B_j = \emptyset \text{ for any } i \neq j$$

mutually disjoint

The Sample Space

 $\Omega =$ a set of possible outcomes of some random "Omega" (not deterministic) experiment

The list defining the sample space must be:
 Mutually exclusive: Each outcome of the experiment is represented once in the sample space.

Collectively exhaustive: All outcomes are elements of the sample space.

An art: Choosing the "right" granularity, to capture the phenomenon of interest as simply as possible. Modeling in science and engineering involves tradeoffs between accuracy, simplicity, & tractability.



You roll a tetrahedral (4-sided) die 2 times.



Formally, sample space is a set of 4²=16 discrete outcomes
 Can also model outcome via *tree-based sequential description*

Alternative Samples Spaces

You roll a tetrahedral (4-sided) die 2 times.

 $\Omega = \left\{ (x, y) \mid x \in \{1, 2, 3, 4\}, y \in \{1, 2, 3, 4\} \right\}$

 $\Omega_{sum} = \{x + y \mid x \in \{1, 2, 3, 4\}, y \in \{1, 2, 3, 4\}$



- Sample space Ω is a set of 4²=16 discrete outcomes
 Sample space Ω_{sum} = {2,3,4,5,6,7,8} is a set of 7 discrete outcomes
- The two sample spaces represent the same experiment
 Elements in \$\Omega_{sum}\$ don't have the same "probability"

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The Discrete Uniform Law (Combinatorics)

Formalizes the idea of "uniformly random" sampling.

- Let all outcomes be equally likely
- Then,

 $\mathbf{P}(A) = \frac{\text{number of elements of } A}{\text{total number of sample points}}$

- Computing probabilities \equiv counting
- Defines fair coins, fair dice, well-shuffled decks

Uniform Law for a Finite Sample Space

You roll a tetrahedral (4-sided) die 2 times.

X = First roll



4

1,3 1,4

- Let every possible outcome have probability 1/16
 - P((X,Y) is (1,1) or (1,2)) =
 - $P({X = 1}) =$
 - P(X + Y is odd) =
 - $P(\min(X,Y)=2) =$

The Basic Counting Principle

Consider a process that consists of r stages. Suppose that:

- (a) There are n_1 possible results for the first stage.
- (b) For every possible result of the first stage, there are n_2 possible results at the second stage.
- (c) More generally, for all possible results of the first i 1 stages, there are n_i possible results at the *i*th stage.

Then, the total number of possible results of the r-stage process is

Simple examples:

- Number of license plates with 3 letters and 4 digits =
- ... if repetition is prohibited =



The set of choices at each stage can depend on previous choices, as long as the number of choices at each stage is constant.

Permutations and Subsets

• **Permutations:** Number of ways of ordering *n* elements is:

$$n(n-1)(n-2)\cdots 1 = \prod_{i=1}^{n} i = n!$$

• Number of subsets of $\{1, \ldots, n\}$ =

$$\underbrace{2 \cdot 2 \cdots 2}_{n \text{ times}} = 2^n$$



Combinations

- $\binom{n}{k}$: number of *k*-element subsets of a given *n*-element set
- Two ways of constructing an ordered sequence of k distinct items:
- Choose the k items one at a time: $n(n-1)\cdots(n-k+1) = \frac{n!}{(n-k)!}$ choices
- Choose k items, then order them
 (k! possible orders)
- Hence:

$$\binom{n}{k} \cdot k! = \frac{n!}{(n-k)!}$$
$$\binom{n}{k} = \frac{n!}{k!(n-k)!}$$



• The total number of subsets:

$$\sum_{k=0}^{n} \binom{n}{k} = 2^{n}$$

Pascal's Triangle



$$\binom{n}{k} = \frac{n!}{k!(n-k)!}$$

$$(a+b)^n = \sum_{k=0}^n \binom{n}{k} a^{n-k} b^k$$

$$\begin{pmatrix} 0 \\ 0 \end{pmatrix} \\ \begin{pmatrix} 1 \\ 0 \end{pmatrix} \begin{pmatrix} 1 \\ 1 \end{pmatrix} \\ \begin{pmatrix} 2 \\ 0 \end{pmatrix} \begin{pmatrix} 2 \\ 1 \end{pmatrix} \begin{pmatrix} 2 \\ 2 \end{pmatrix} \\ \begin{pmatrix} 3 \\ 0 \end{pmatrix} \begin{pmatrix} 3 \\ 1 \end{pmatrix} \begin{pmatrix} 3 \\ 2 \end{pmatrix} \begin{pmatrix} 3 \\ 2 \end{pmatrix} \begin{pmatrix} 3 \\ 3 \end{pmatrix} \\ \begin{pmatrix} 4 \\ 1 \end{pmatrix} \begin{pmatrix} 4 \\ 2 \end{pmatrix} \begin{pmatrix} 4 \\ 3 \end{pmatrix} \begin{pmatrix} 4 \\ 4 \end{pmatrix}$$

$$\sum_{k=0}^{n} \binom{n}{k} = 2^{n}$$

Pascal's Triangle



http://www.mathwarehouse.com/

Binomial Probabilities

If I toss a coin *n* times, what is the probability that I see *k* heads?

- *n* independent coin tosses
 - $\mathbf{P}(H) = p = \frac{1}{2}$
- P(HTTHHH) =



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 Probability: Counting and the Discrete Uniform Law
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Suppose there are *m* students in a class. What is the probability that at least two students in the class have the same birthday?

In a class of 70 students, this probability is about 99.87%

Assumptions:

Birthdays are equally likely to occur on any of N=365 days
 No dependence between birthdays of different students
 Not completely true, but fairly accurate approximations.

 N possible birthdays, m students in the class.
 We can compute this probability by counting elements of the sample space of all N^m possible birthday patterns:

$$\Omega = \{ (b_1, \dots, b_m) \mid b_i \in \{1, \dots, N\} \}$$

➤ The number of birthday patterns with all pairs distinct: $D_m = \{(b_1, \ldots, b_m) \mid b_i \neq b_j \text{ for all } i \neq j\}$ $|D_m| = N(N-1)(N-2)\cdots(N-m+1) = \frac{N!}{(N-m)!}$

Sample Space: $\Omega = \{(b_1, ..., b_m) \mid b_i \in \{1, ..., N\}\}$

Set of items in the sample space with all distinct birthdays:

$$D_m = \{(b_1, \ldots, b_m) \mid b_i \neq b_j \text{ for all } i \neq j\}$$

In the uniform probability model:

$$\frac{|Dm|}{|\Omega|} = P(D_m) = \frac{N!}{(N-m)!N^m} = \prod_{i=0}^{m-1} \left(\frac{N-i}{N}\right)$$

- Sort students in arbitrary order (say, alphabetical by name)
 Define two events for student *j* in the list of *m* students:
 - $R_j \rightarrow$ birthday j is a repeat of some previous student $D_j \rightarrow$ all of the first j birthdays are distinct

> Tree diagram of event probabilities for N=365 days:



- The probability that the first *m* birthdays are distinct is then: $P(D_m) = \left(1 - \frac{1}{N}\right) \left(1 - \frac{2}{N}\right) \cdots \left(1 - \frac{m-1}{N}\right) = \prod_{i=0}^{m-1} \left(1 - \frac{i}{N}\right)$
 - $R_j \rightarrow$ birthday *j* is a repeat of some previous student $D_j \rightarrow$ all of the first *j* birthdays are distinct
- > Tree diagram of event probabilities for N=365 days:



 \succ The probability that the *m* birthdays are distinct:

$$P(D_m) = \frac{N!}{(N-m)!N^m} = \prod_{i=0}^{m-1} \left(\frac{N-i}{N}\right)$$

$$P(D_m) = \left(1 - \frac{1}{N}\right) \left(1 - \frac{2}{N}\right) \cdots \left(1 - \frac{m-1}{N}\right) = \prod_{i=0}^{m-1} \left(1 - \frac{i}{N}\right)$$

 \succ The probability that *m* birthdays on *N* days are distinct:

(curve) $P(D_m) \leq \prod_{i=0}^{m-1} e^{-\frac{i}{N}} = e^{-\sum_{i=0}^{m-1} \frac{i}{N}} = e^{-\frac{m(m-1)}{2N}}$

 \succ The probability that *m* birthdays on *N* days are distinct:



Probability that first repeated birthday is found with student *j*

Countable and Uncountable Infinite

Advanced topic not covered in homeworks or exams!

- A finite set is countable
- The set of natural numbers N = {1, 2, 3, ...} is countable
- A set S is countable if there is injective mapping from S to N
- The set of all integers is countable 1, -1, 2, -2, 3, -3,....
- The set of all rationales is countable



The Rational numbers in [0,1] is Countable

Advanced topic not covered in homeworks or exams!

- Each rational number $\frac{a}{b}$ in [0,1] has a place in the table
- There is an order in the table that associates each rational number with a natural number.



Source: www.homeschoolmath.net/teaching/rational-numbers-countable.php

The Real Numbers in [0,1] are Uncountable

Advanced topic not covered in homeworks or exams!

Cantor's diagonal argument

- Assume that the real numbers in [0,1] were countable.
- Write each real number in [0,1] in binary fraction
- If countable can be ordered in a list
- Construct a real number s such that its i-th bit is the complement the i-th bit of the i-th number in the list.
- The number s doesn't appear in the list.
 For i>0 the number s is different from the ith number in its i-th bit.

 $s_1 = 0 0 0 0 0 0 0 0 0 0 0 0 0 \dots$ $s_3 = 0 \ 1 \ 0 \ 1 \ 0 \ 1 \ 0 \ 1 \ 0 \ 1 \ 0 \ \ldots$ $s_4 = 1 \ 0 \ 1 \ 0 \ 1 \ 0 \ 1 \ 0 \ 1 \ 0 \ 1 \ \dots$ $s_5 = 1 \ 1 \ 0 \ 1 \ 0 \ 1 \ 0 \ 1 \ 0 \ 1 \ \dots$ $s_6 = 0\ 0\ 1\ 1\ 0\ 1\ 1\ 0\ 1\ 1\ 0$... $s_7 = 1\ 0\ 0\ 0\ 1\ 0\ 0\ 1\ 0\ 0$... $s_8 = 0\ 0\ 1\ 1\ 0\ 0\ 1\ 1\ 0\ 0\ 1$... $s_{10} = 1\ 1\ 0\ 1\ 1\ 1\ 0\ 0\ 1\ 0\ 1\ \ldots$ $s_{11} = 1\ 1\ 0\ 1\ 0\ 1\ 0\ 0\ 1\ 0\ 0$...

https://en.wikipedia.org/wiki/Cantor%27s_diagonal_argument

Discrete vs. Continuous Spaces

- Countable sample space → discrete probability model
- Uncountable sample space → continuous probability model