## Probability 1: The Basic

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# Probability



What is a reasonably safe gathering size in the age of COVID?

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  - The heart of statistical inference

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Linda is 31 years old, single, outspoken, and very bright. She majored in philosophy. As a student, she was deeply concerned with issues of discrimination and social justice, and also participated in anti-nuclear demonstrations

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Famous example of the conjunction fallacy called the Linda problem
 Majority of respondents chose 2, but this is impossible!



• Learning mathematical probability avoids these mistakes!

# Learning about populations



- **Probability**: formalize the uncertainty about how our data came to be.
- Inference: learning about the population from a set of data.

#### Roadmap

- Naive Definition of Probability
- Non-naive Definition of Probability

1. Naive Definition of Probability

## Sample spaces & events



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  - ▶ To formalize, we need to define the set of possible outcomes.
- A sample space  $\Omega$  is the set of possible outcomes.
  - Can be finite, countably infinite, or uncountably infinite.
- $\omega \in \Omega$  is one particular outcome.
- A subset of  $\Omega$  is an event and we write this as  $A\subset \Omega$

# Naive definition of probability

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- Often wrong, why?
- Often wrong, but justified under a few assumptions:
  - **Symmetry**: fair coin, shuffled cards, dice, etc.
  - **Simple random sample** from a population.
- Example: randomly select a card from a standard deck of playing cards
  - Each of the 52 card has equal probability

$$P(10\clubsuit) = P(8\heartsuit) = \frac{1}{52}$$

# Counting

- **Multiplication rule**: if you have two sub-experiments, *A* with *a* possible outcomes and *B* with *b* possible outcomes, then in the combined experiment there are *ab* possible outcomes.
- Example: what to watch and where to watch?
  - What to watch? Netflix or Hulu.
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- Example: what to watch and where to watch?
  - ▶ What to watch? Netflix or Hulu.
  - ▶ Where to watch? TV, tablet, or phone.
  - $2 \times 3 = 6$  possible outcomes (Netflix on TV, Hulu on phone, ...)
- Assumes number of outcomes in one experiment *does not* depend on the outcome of the other experiment  $\rightarrow$  big assumption, often wrong in real world

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- Ex: 2024 best grad student paper awards has 11 candidates. How many top-3 award possibilities?
  - 11 first-place choices
  - ▶ 10 second-place choices among the remaining candidates
  - 9 third-place choices
  - Total:  $11 \cdot 10 \cdot 9 = 990$  possibilities

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2. Non-Naive Definition of Probability

# Probability

- A probability space consists of:
  - Sample space  $\Omega$
  - Probability function  $\mathbb{P}$  mapping events  $\mathsf{A} \subseteq \Omega$  onto the real line.
- The function  $\mathbb{P}$  must satisfy the following axioms:
  - (Non-negativity)  $\mathbb{P}(A) \ge 0$  for every event
  - **2** (Normalization)  $\mathbb{P}(\Omega) = 1$
  - (Additivity) If a series of events,  $A_1, A_2, ..., are disjoint$ , then
    - Events A and B are disjoint if their intersection is zero:  $P(A \cap B) = 0$
    - E.g. being early to class and being late to the same class

$$\mathbb{P}(\cup_{n=1}^{\infty}A_i)=\sum_{i=1}^{\infty}\mathbb{P}(A_i)$$

• Union of mutually exclusive events  $\rightsquigarrow$  use additivity

14/17

### Interpretation of probabilities

- $\bullet$  How do we interpret  $\mathbb{P}(A)?$  Huge debate about this in stats literature
  - **I** Frequentist:  $\mathbb{P}()$  reflect relative frequency in a large number of trials.
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  - Repeat a coin flip many times frequency of head  $\approx 0.5$
- **2** Bayesian:  $\mathbb{P}()$  are subjective beliefs about outcomes
  - How likely I think a particular event will be
- Both viewpoints are helpful in different contexts
  - Properties of probabilities exactly the same in either approach
  - This class: focus on frequentist perspectives. But we will talk a bit about Bayes next week, so you know how it works given that it's becoming more popular.

## Gambling 101

- What's the probability of selecting a 8 card from a well-shuffled deck?
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# Gambling 101

- What's the probability of selecting a 8 card from a well-shuffled deck?
  "Well-shuffled" → "randomly selected" → all cards have a chance
- "8 card" event =  $\{8 \clubsuit \cup 8 \diamondsuit \cup 8 \heartsuit \cup 8 \diamondsuit \}$
- Union of mutually exclusive events ~> use additivity

$$\mathbb{P}$$
 (a 8 card) =  $\mathbb{P}(8\clubsuit) + \mathbb{P}(8\spadesuit) + \mathbb{P}(8\heartsuit) + \mathbb{P}(8\diamondsuit) = \frac{4}{52}$ 

Some properties of probabilities (refresher)

- Probability of not A is 1 minus the probability of A.
- Follows from  $\mathsf{A}\cup\mathsf{A}'=\Omega$  and  $1=\mathbb{P}(\Omega)=\mathbb{P}(\mathsf{A})+\mathbb{P}(\mathsf{A}')$

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- Probability of not A is 1 minus the probability of A.
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- $\ \ \, {\hbox{\rm lf}} \ \, A\subset B, \ {\hbox{\rm then}} \ \, \mathbb{P}(A)\leq \mathbb{P}(B) \\$ 
  - Subsets of events have lower probability than the event.

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2 If 
$$A \subset B$$
, then  $\mathbb{P}(A) \leq \mathbb{P}(B)$ 

• Subsets of events have lower probability than the event.

- Avoid "double-counting" the part where A and B overlap.
- Inclusion-exclusion