

# 1: Basic Probability

Spring 2021

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Gov 2002 (Harvard)



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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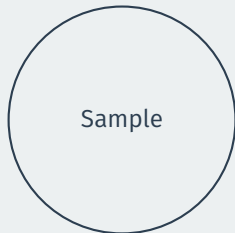
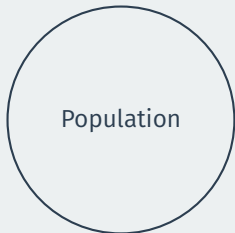
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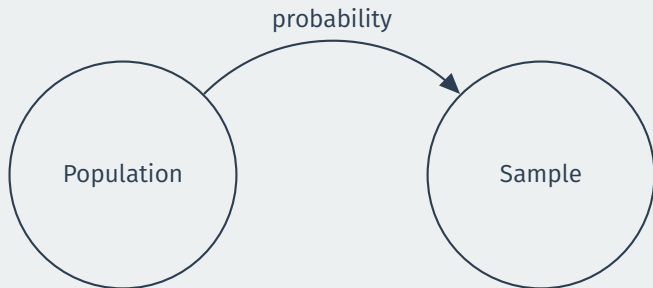
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- Learning mathematical probability avoids these mistakes!

# Learning about populations

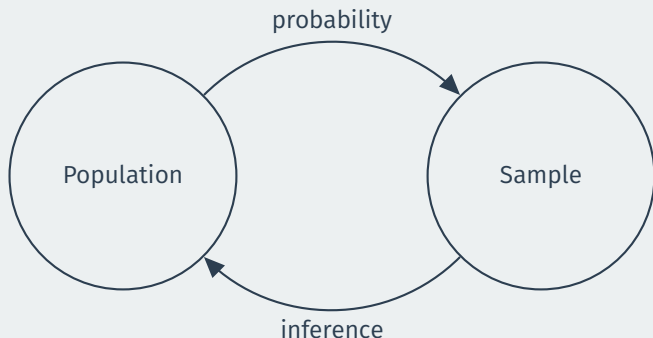


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- **Probability:** formalize the uncertainty about how our data came to be.
- **Inference:** learning about the population from a set of data.

# Roadmap

1. Naive Definition of Probability
2. Non-naive Definition of Probability

# 1/ Naive Definition of Probability

# Sample spaces & events

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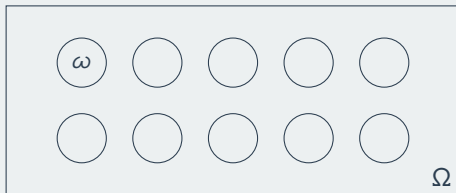
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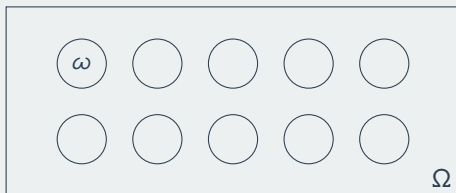
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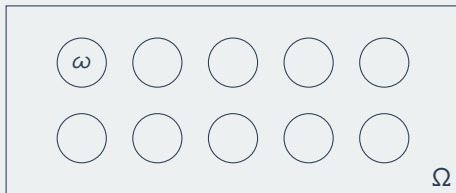
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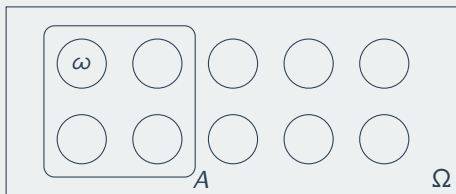
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- $\omega \in \Omega$  is one particular **outcome**.
- A subset of  $\Omega$  is an **event** and we write this as  $A \subset \Omega$ .

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$$\mathbb{P}(4\clubsuit) = \mathbb{P}(4\heartsuit) = 1/52$$

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- Assumes number of outcomes in one experiment doesn't depend on the outcome of the other experiment.

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  - Total:  $11 \cdot 10 \cdot 9 = 990$  possibilities

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- **Binomial coefficient:** number of subsets of size  $k$  in a group of  $n$  objects:

$$\binom{n}{k} = \frac{n(n-1) \cdots (n-k+1)}{k!} = \frac{n!}{(n-k)!k!}$$

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  - Prepare 8 cups of tea, 4 milk-first, 4 tea-first
  - Present cups to friend in a **random** order
  - Ask friend to pick which 4 of the 8 were milk-first.

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- $\rightsquigarrow$  the guessing hypothesis might be implausible.

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There are  $k$  people in a room. Assume each person's birthday is equally likely to be any of the 365 days of the year (no leap babies) and birthdays are independent. What is the probability that at least one pair of people have the same birthday?

## **2/** Non-naive Definition of Probability

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  3. (Additivity) If a series of events,  $A_1, A_2, \dots$ , are disjoint, then

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- Probability function assigns “mass” to regions of the sample space.

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    - This class: focus on frequentist perspectives because it's pervasive.

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- **Inclusion-exclusion**

# Appendix

- A standard deck of playing cards has 52 cards:

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- An event: picking a Queen, {Q♣, Q♠, Q♥, Q♦}

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