

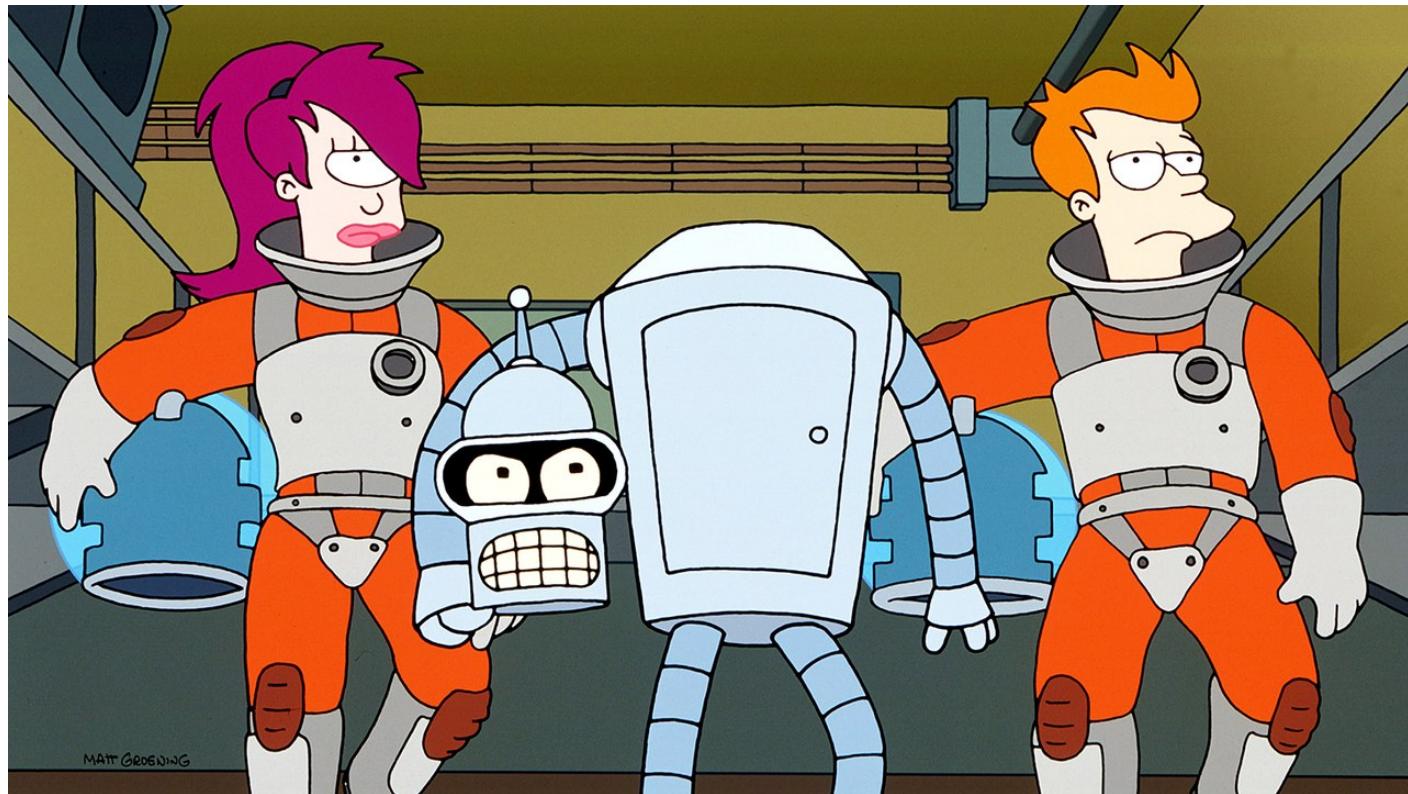
# **Math Camp: Lesson 4**

## **Statistics and Probability**

**UW–Madison Political Science**

**September 2, 2021**

Hang in there



# Agenda

- Why do we need statistics?
- Counting
- Set theory
- Probability
- Independence, joint probability
- Bayes' Theorem
- Looking ahead

# Why statistics?

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- We build a model that includes a number of variables (e.g., economy, conflict, etc.)
- In the model, we have parameters associated with these factors that tell us about their influence
- Statistics allows us:
  - to estimate these parameters, to learn which factors are systematically related to our outcome of interest
  - to estimate how (un)certain we should be in our estimates

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In political science, we do both, but a lot of emphasis is on inference and causality

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- We entertain different **models** for the process and pick the best model based on the probability of data under each model
- Probability also helps us evaluate the level of uncertainty around our findings

# Helpful vocabulary

A **random variable** is a realization of a process that is at least partially random (i.e. unpredictable)

- e.g. coin flip, dice roll, regime failure (or absence of failure)
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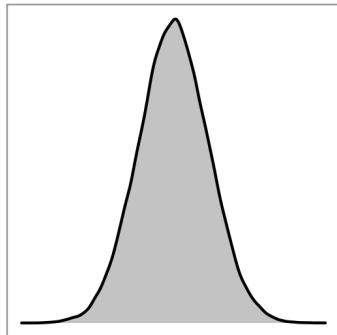
The probability of all potential outcomes is described with a **probability distribution**, a *function* that maps each potential outcome to a certain probability

- $x$  = potential outcome
- $f(x)$  = probability of  $x$

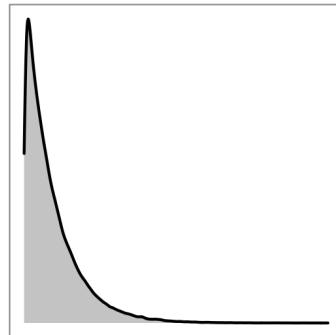
These also matter for formal (non-statistical) models (e.g., an actor learns certain information with some probability)

# Common probability distributions

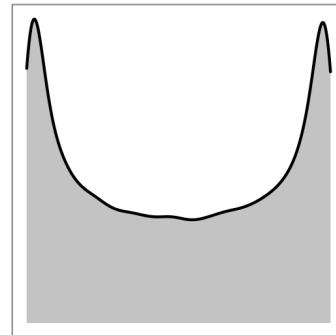
Normal



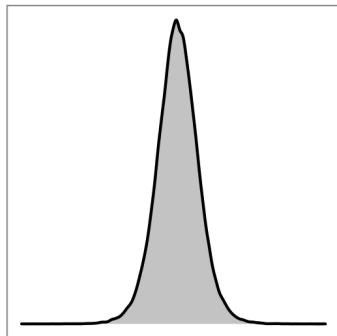
Poisson



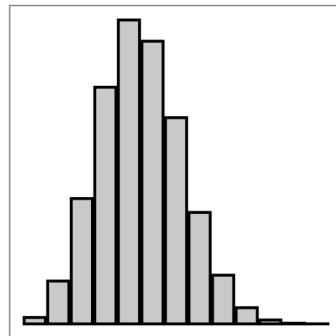
Beta (0.5, 0.5)



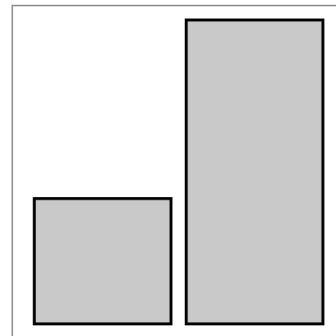
T (df = 10)



Binomial



Bernoulli



Based on simulated data

# **How probability distributions work**

Probability distributions can describe discrete outcomes (regime survival or failure) or continuous outcomes (election turnout, vote margin)

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Probability distributions always sum to 1

- "The law of total probability" (allows to calculate probabilities of certain events based on probabilities of other events)
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Probability distributions are the basis for statistical inference

- $z$ -scores,  $p$ -values
- Prior and posterior beliefs

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Suppose an **event** is described by  $K$  different component parts. (E.g., we roll a die  $K$  many times.) Each component  $k = \{1, 2, \dots, K\}$  has  $n_k$  possible values. What is the number of distinct outcomes we could get?

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I roll a 6-sided die 4 times. How many unique sets of 4 rolls can I obtain (assuming that different orderings of the same 4 numbers are different events)?

# Complex counting considerations

Does the *order of selection* matter? (Is  $\{1, 2\} = \{2, 1\}$ ?)

Are selected objects *replaced* (able to be selected again) or *not replaced*?

(You might hear the terms "combination" and "permutation". Usually, "permutation" refers to situations where order matters, and "combination" refers to situations where order does not matter.)

# Ordered, with replacement

This is easiest because (a) no need to adjust for "double-counting" and (b) the number of possibilities is always constant.

The number of possible ways to select  $k$  elements from a larger pool of  $n$  is

$$n \times n \times n \times \dots \times n = n^k$$

Intuition: in each draw, there are  $n$  possibilities. Each of  $n$  outcomes in one draw can be combined with the  $n$  outcomes in any (and all) other draws.

Examples: rolling two dice several times, PIN codes, phone numbers

# Ordered, without replacement

The number of ways to select  $k$  objects from a pool of  $n$  possible objects, where order matters, but replacement does not occur.

$$n * (n - 1) * (n - 2) * \dots * (n - k - 1) = \frac{n!}{(n - k)!}$$

Intuition: each draw *removes an object* from the larger pool. Subsequent draws have one less element to choose from.

Example: winning lottery numbers, ranking of candidates in an election

# Unordered, without replacement

The number of ways to select  $k$  objects from a pool of  $n$  possible objects, where order does not matter and replacement does not occur.

Intuition: we have fewer possibilities than before. Substantively identical elements ( $A$  and then  $B$ ;  $B$  and then  $A$ ) are not double counted

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

For example: survey samples, raffles, possible groups of 2 in a classroom

# Unordered, with replacement

The number of possible ways to select  $k$  elements from a larger pool of  $n$  possible elements, where order does not matter and replacement does occur

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

Examples: the number of heads if you flip a coin  $n$  times

# Exercises

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Imagine we have 2 identical bicycles for students in this class. You can only win 1 bicycle. How many sets of winners?

# Set Theory

# Sets

Remember: a **set** is a collection of elements. Could be numbers, units, areas in space...

- $F = \{1, 2, 3, 4\}$
- $G = \{1, 3, 5\}$
- $H = [0, 1] \cup (2, 3)$

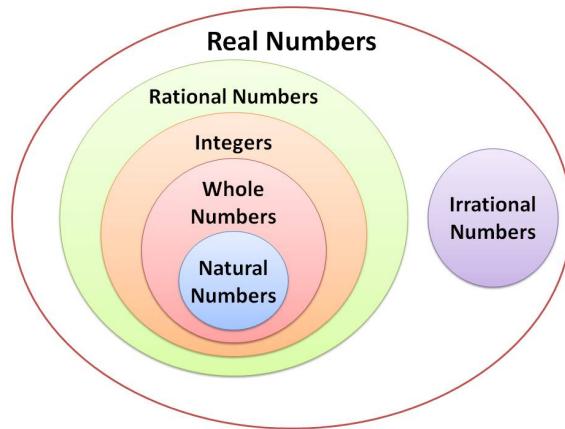
What are unions?

Intersections?

Disjoints?

Subsets?

Supersets?



# The sample space

The **sample space** (denoted  $S$  or  $\Omega$ ) is the set that contains all the elements in question.

Sometimes called the *universal set*.

"Universal" does not mean that it contains *everything*. Rather, it's all the relevant objects: e.g., the universal set for a nationally representative survey sample is a country's population (not all the people in the world).

# Complementary sets

The **complement** of set  $A$  (denoted as  $A^C$ ) is the set of all elements in the sample space that are *not contained* in  $A$

$$A^C \equiv X \text{ such that } X \notin A$$

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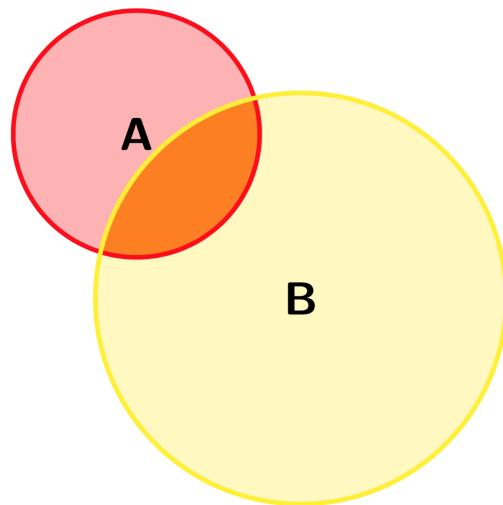
What is  $\Omega^C$ ?

- $\emptyset$

# Probability

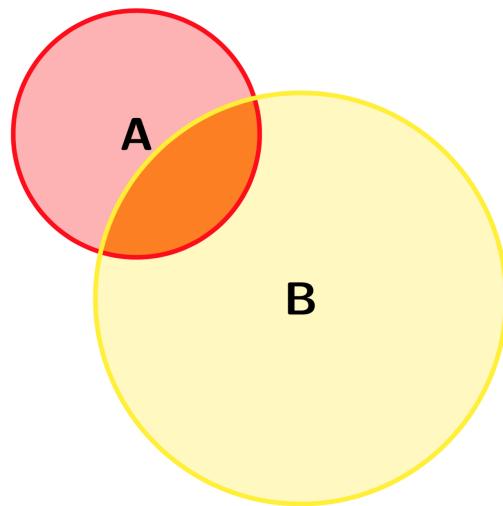
# Sets and probability

We can use sets to represent the probability of events. Total area represents the total probability of all events (equal to 1).



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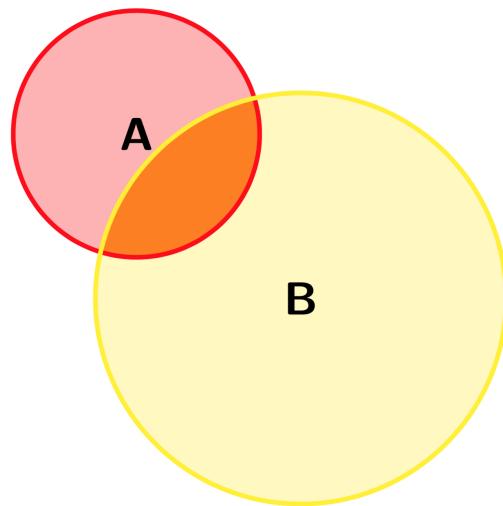
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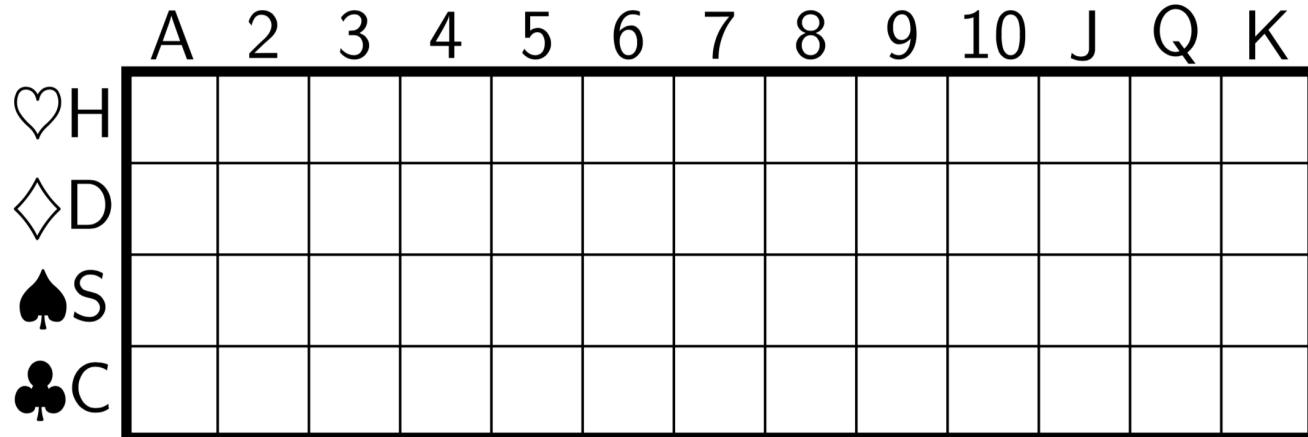
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# Let's play cards

There are 4 suits (hearts, diamonds, spades, clubs) and 13 card values (Ace, 2, 3, ..., Jack, Queen, King). Suits and values can both be sets.

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Total area = 1

Probability of an individual card:  $\frac{1}{52}$

# Properties of probabilities

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If we have  $N$  many *collectively exhaustive* and *mutually exclusive* sets of potential outcomes, their probabilities sum to 1. Which is to say, *something must happen*.

$$\sum_{n=1}^N p(A_n) = 1$$

# Probability of complements

If  $\Omega$  contains the set of all potential outcomes, and  $A$  is an event that is a subset of the outcome space that occurs with  $p(A)$ ...

What is  $p(A^C)$ ?

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The intuition: *Something* must happen, either  $A$  or not  $A$

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Probability that a random card is a Heart?  $p(H) = \frac{1}{4}$

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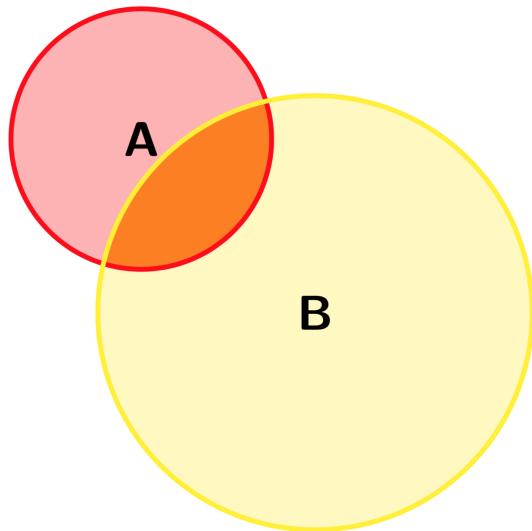
	A	2	3	4	5	6	7	8	9	10	J	Q	K
♡H													
◇D													
♠S													
♣C													

Probability that a card is not a Heart?  $1 - p(H) = \frac{3}{4}$

# Probability of unions

The probability of  $A \cup B$

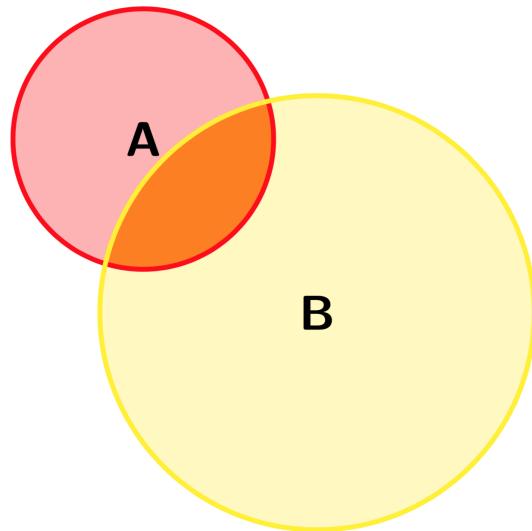
The probability that *either*  $A$  or  $B$  occurs



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$$p(A \cup B) = p(A) + p(B) - p(A \cap B)$$

The intuition: the sum of  $A$  and  $B$  will double count  $A \cap B$ , so we need to subtract one instance of  $A \cap B$

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What is the probability that we draw a card that is *either* a Heart *or* a face card?

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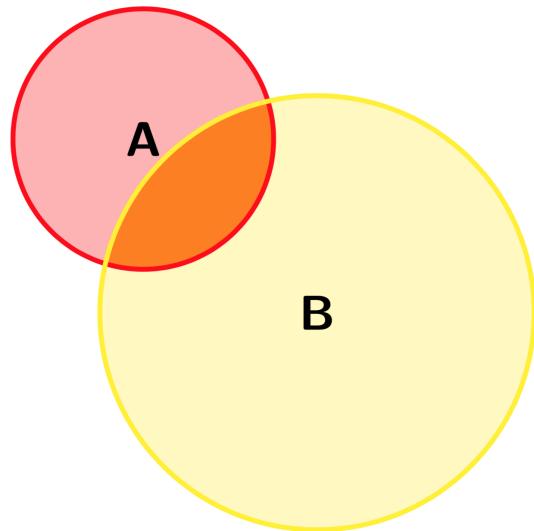
$$p(H \cap F) = ?$$

$$p(H \cup F) = \frac{1}{4} + \frac{12}{52} - \frac{3}{52} = \frac{22}{52}$$

# Probability of intersections

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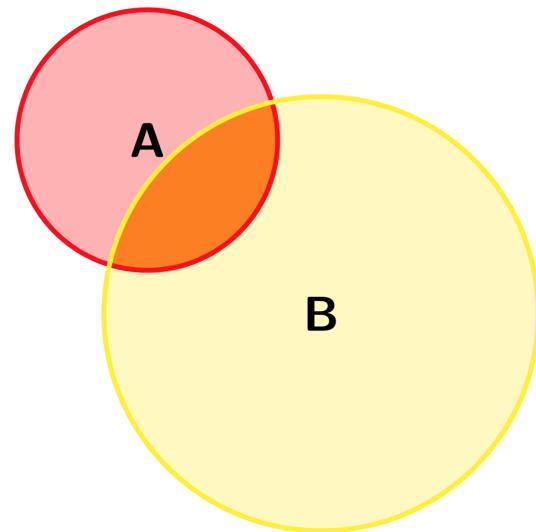
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$$p(A \cap B) = p(A) + p(B) - p(A \cup B)$$

The intuition: we care only about the component that we double counted

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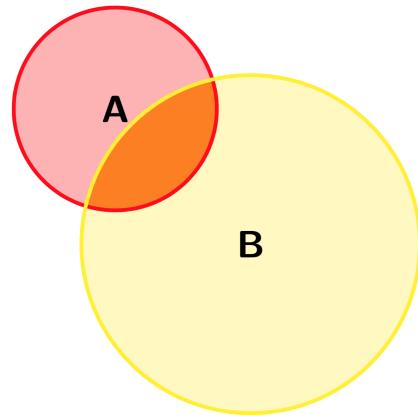
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$$p(H \cap F) = \frac{1}{4} + \frac{12}{52} - \frac{22}{52} = \frac{3}{52}$$

# Conditional probability

The probability of  $A$ , given  $B$ , is expressed as  $p(A | B)$

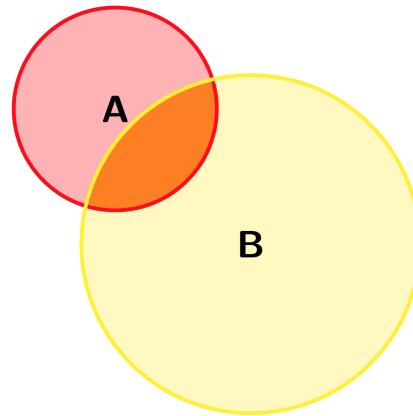
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$$p(A | B) = \frac{p(A \cap B)}{p(B)}$$

The intuition:

- If we *know* that  $B$  happened, we only care about the space within  $B$
- The probability that both  $A$  and  $B$  happen, divided by the probability of  $B$
- $p(\text{intersection}) / p(\text{conditioning event})$

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	A	2	3	4	5	6	7	8	9	10	J	Q	K
♥ H													
♦ D													
♠ S													
♣ C													

What is the probability of drawing the Ace of Diamonds?

# Conditional probability

	A	2	3	4	5	6	7	8	9	10	J	Q	K
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What is the probability of drawing the Ace of Diamonds?

What is the probability of drawing the Ace of Diamonds, *given that* we have drawn an Ace?

# Conditional probability

	A	2	3	4	5	6	7	8	9	10	J	Q	K
♥ H													
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What is the probability of drawing the Ace of Diamonds?

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- $p(\text{Ace of Diamonds}) = \frac{1}{52}$

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# Conditional probability

	A	2	3	4	5	6	7	8	9	10	J	Q	K
♥ H													
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What is the probability of drawing the Ace of Diamonds?

What is the probability of drawing the Ace of Diamonds, *given that* we have drawn an Ace?

- $p(\text{Ace of Diamonds}) = \frac{1}{52}$
- $p(\text{Ace}) = \frac{4}{52}$
- $p(\text{Ace of Diamonds} \mid \text{Ace}) = \frac{1/52}{4/52} = \frac{1}{4}$

# Conditional probability in political science

Many questions we ask in political science are about conditional probabilities. For example:

- What is the probability that **an authoritarian government increases welfare spending**, given that it performs poorly in an election?
- How likely is **a police officer to shoot a minority civilian**, given that the officer is white/non-white?
- What is the probability that **a respondent expresses positive feelings toward immigrants** conditional on the randomly assigned treatment—learning about their family's immigration history?
- How likely **is an incumbent governor to be re-elected** if unemployment crosses a certain threshold?

# Exercises: what's the probability?

	A	2	3	4	5	6	7	8	9	10	J	Q	K
♥H													
♦D													
♠S													
♣C													

$$p(\{8, 9, 10\})$$

$$p(\{5, 6\} \cup \{6, 10\})$$

$$p(A \mid H^C)$$

# Exercises: what's the probability?

We have a sample of democratic regimes, some of which have broken down (note: the numbers are completely fictional):

Country wealth	Breakdown	No breakdown
Poor	23	56
Wealthy	5	115

What is the probability that a randomly chosen democracy from this sample is poor and has broken down?

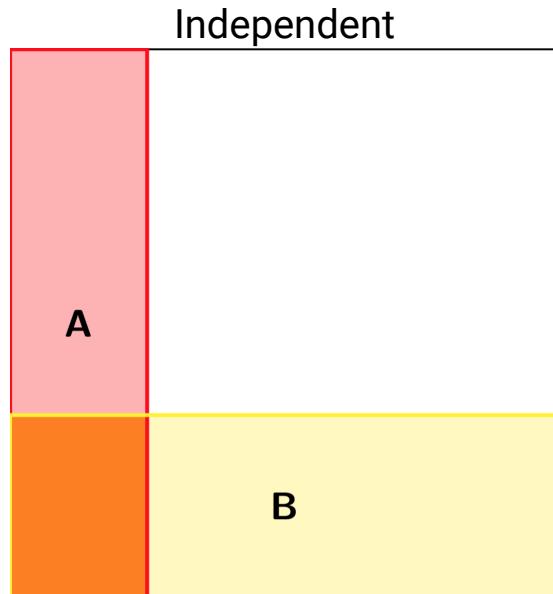
What is the probability that a randomly chosen democracy breaks down, given that it is wealthy?

# The notion of *independence*

Two events are **independent** if knowing the outcome of one event does not change the probability of the other.

# The notion of *independence*

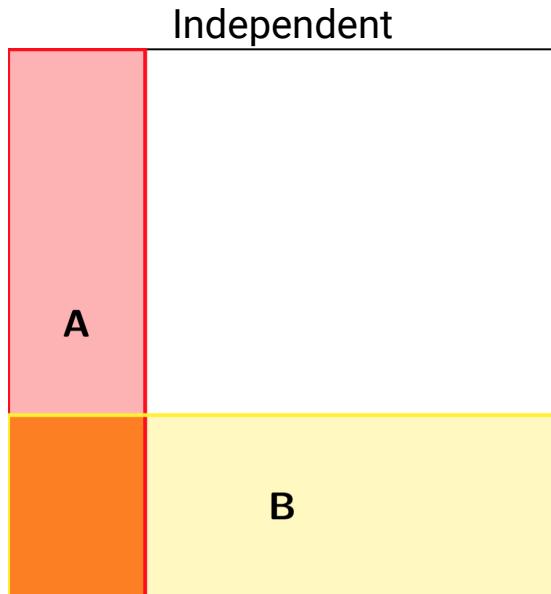
Two events are **independent** if knowing the outcome of one event does not change the probability of the other.



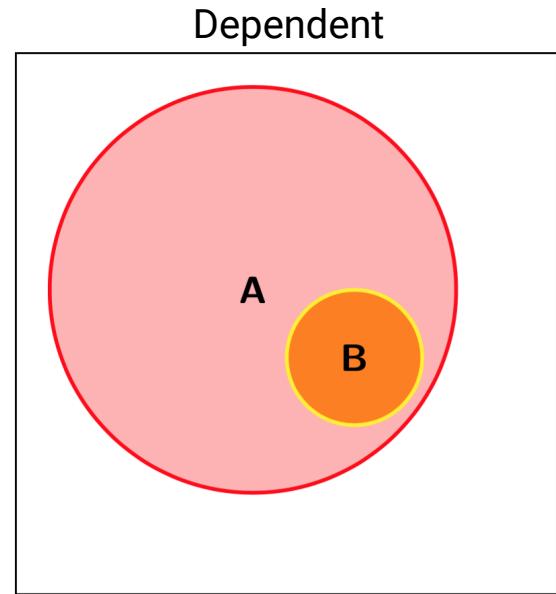
$$p(B) = p(B \mid A)$$

# The notion of *independence*

Two events are **independent** if knowing the outcome of one event does not change the probability of the other.



$$p(B) = p(B \mid A)$$



$$p(B) \neq p(B \mid A)$$

# Independence of events

Is drawing a face card independent of drawing a Hearts card?

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♥H													
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$$p(F \mid H) = \frac{3}{13}$$

# Independence of events

Is drawing a face card independent of drawing a Hearts card?

	A	2	3	4	5	6	7	8	9	10	J	Q	K
♥ H													
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♠ S													
♣ C													

$$p(F \mid H) = \frac{3}{13}$$

$$p(F) = \frac{12}{52} = \frac{3}{13}$$

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What about drawing a face card independent of drawing a card greater than 8?

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$$p(X = F \mid X > 8) = \frac{12}{20} = \frac{3}{5}$$

# Independence of events

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$$p(X = F \mid X > 8) = \frac{12}{20} = \frac{3}{5}$$

$$p(F) = \frac{12}{52} = \frac{3}{13}$$

# Joint probability

What we're doing here is considering the probability of *multiple events*.

**Joint probability**: the probability of *more than one event* occurring simultaneously.

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**Joint probability**: the probability of *more than one event* occurring simultaneously.

$$p(A, B) \equiv p(A) \cap p(B)$$

# Joint probability of independent events

If multiple events are independent of one another, the joint probability of all events is the *product* of the individual probabilities.

Example: we flip three coins independently of one another. What's the probability of the sequence  $\{H, H, H\}$ ?

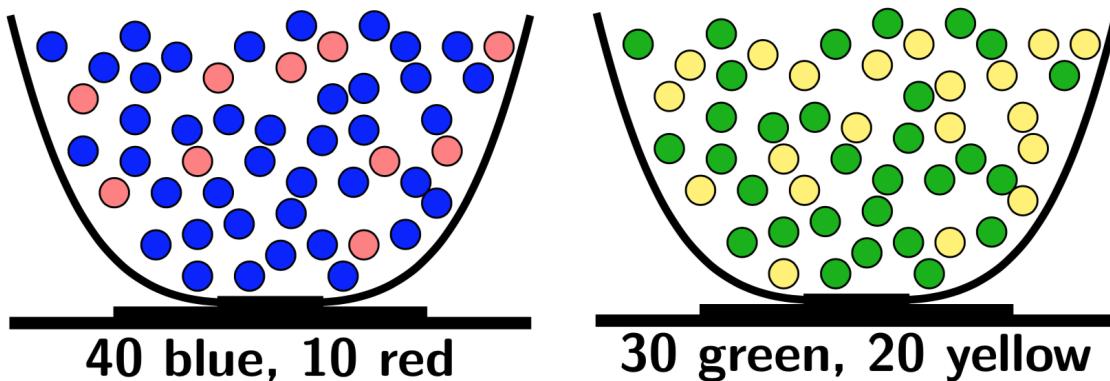
# Joint probability of independent events

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Example: we flip three coins independently of one another. What's the probability of the sequence  $\{H, H, H\}$ ?

$$\begin{aligned} p(H) \times p(H) \times p(H) &= .5 \times .5 \times .5 \\ &= 0.125 \end{aligned}$$

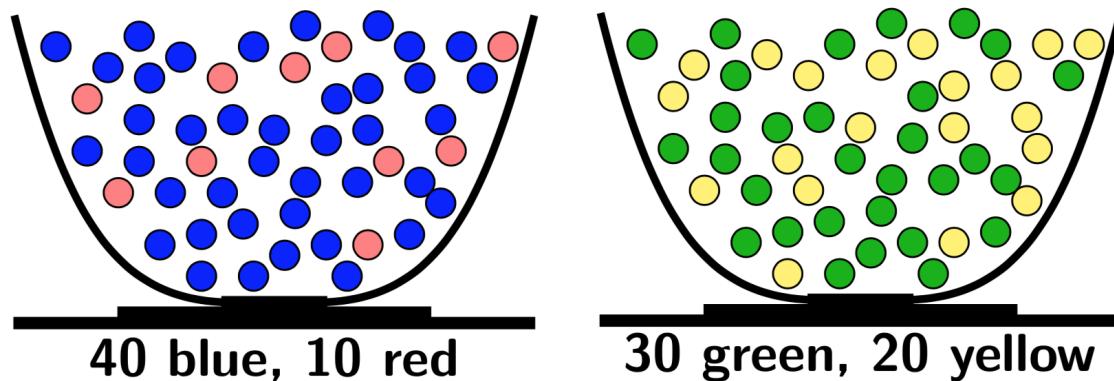
## Joint probability of independent events



We've got two bowls. If we draw a ball from each bowl, what is the joint probability of...

- $p(\text{blue, green}) = ?$
- $p(\text{blue, yellow}) = ?$
- $p(\text{red, green}) = ?$
- $p(\text{red, yellow}) = ?$

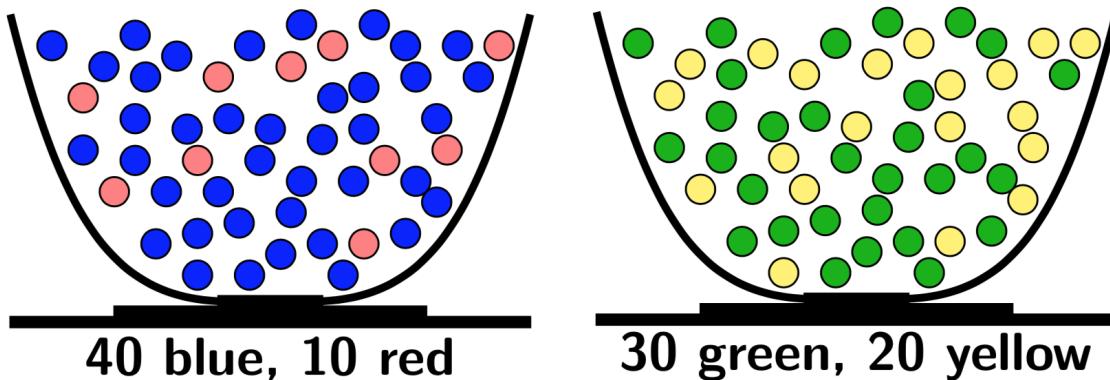
## Joint probability of independent events



We've got two bowls. If we draw a ball from each bowl, what is the joint probability of...

- $p(\text{blue, green}) = \left(\frac{40}{50}\right) \left(\frac{30}{50}\right) = (.8)(.6) = .48$
- $p(\text{blue, yellow}) = \left(\frac{40}{50}\right) \left(\frac{20}{50}\right) = (.8)(.4) = .32$
- $p(\text{red, green}) = \left(\frac{10}{50}\right) \left(\frac{30}{50}\right) = (.2)(.6) = .12$
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Because these are mutually exclusive and exhaustive events, probabilities sum to 1

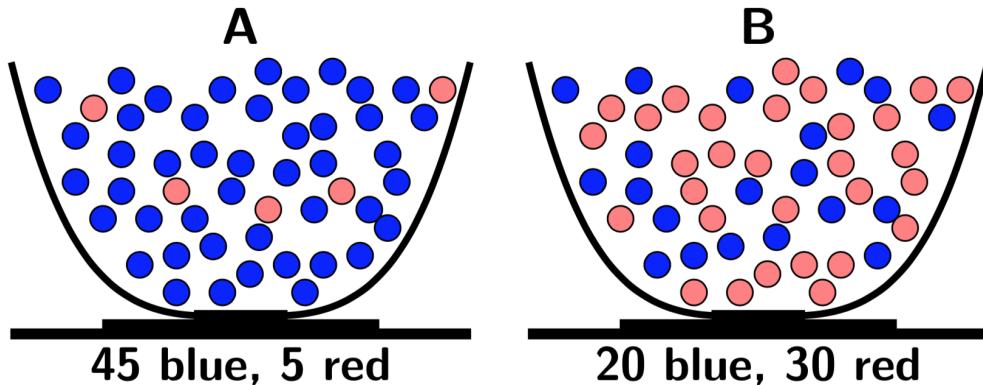
# Exercises

In a given district, 35% support party  $A$ , 50% support party  $B$ , and 15% support party  $C$ . The turnout in the previous local election was 20%, and supporters of all three parties were equally likely to vote. What is the probability that a randomly chosen adult in the district supports party C and has voted in the last election?

In a sample of countries, 45% are democratic and 55% are authoritarian. 10% of the countries, regardless of their political regime, have experienced an oil windfall in the last 20 years. How likely is that a randomly selected country is a democracy and has experienced an oil windfall?

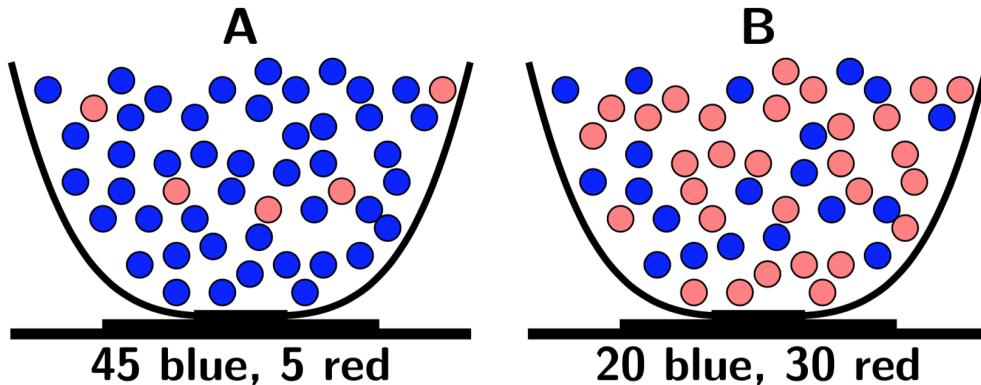
## Conditional and total probability

We flip a coin. If heads, we draw a ball from the left urn. If tails, we draw from the right.



## Conditional and total probability

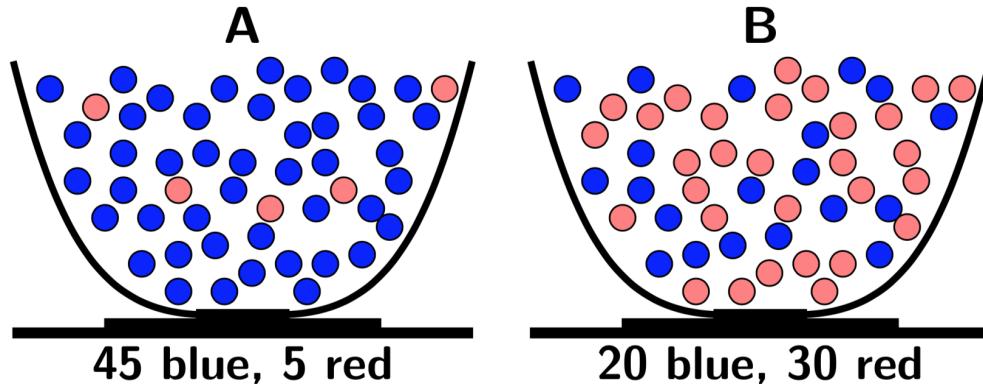
We flip a coin. If heads, we draw a ball from the left urn. If tails, we draw from the right.



This means there are two ways to choose a blue ball:  $\{A, \text{blue}\}$  and  $\{B, \text{blue}\}$

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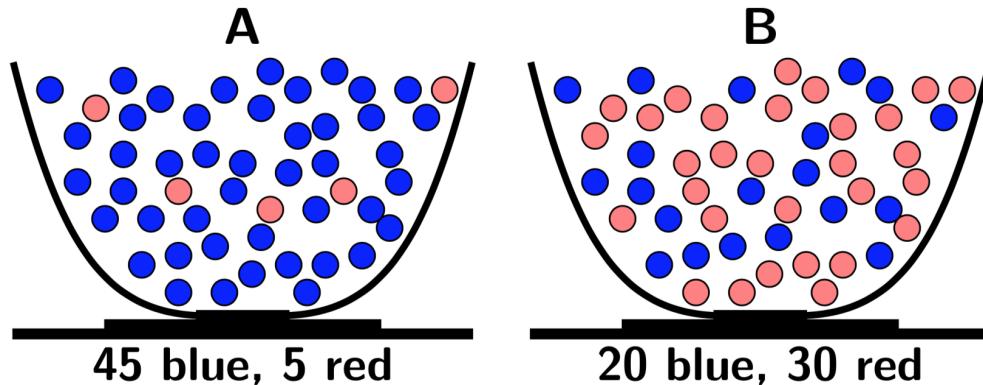


This means there are two ways to choose a blue ball:  $\{A, \text{blue}\}$  and  $\{B, \text{blue}\}$

- $p(A, \text{blue}) = 0.5 * \frac{45}{50} = 0.45$
- $p(B, \text{blue}) = 0.5 * \frac{20}{50} = 0.20$

## Conditional and total probability

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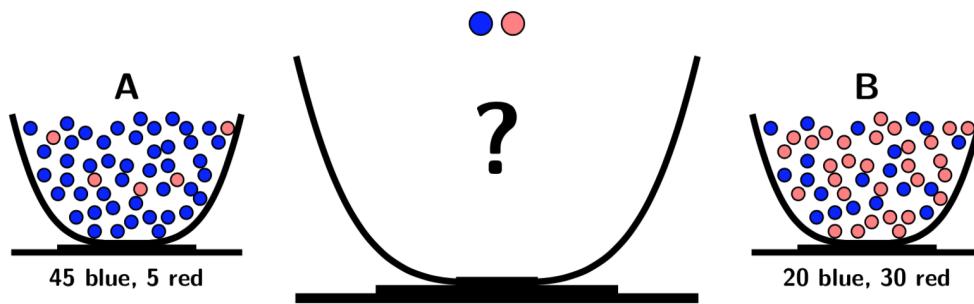
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- $p(B, \text{blue}) = 0.5 * \frac{20}{50} = 0.20$

Total probability of blue is the sum of the joint probabilities (a very useful principle...)

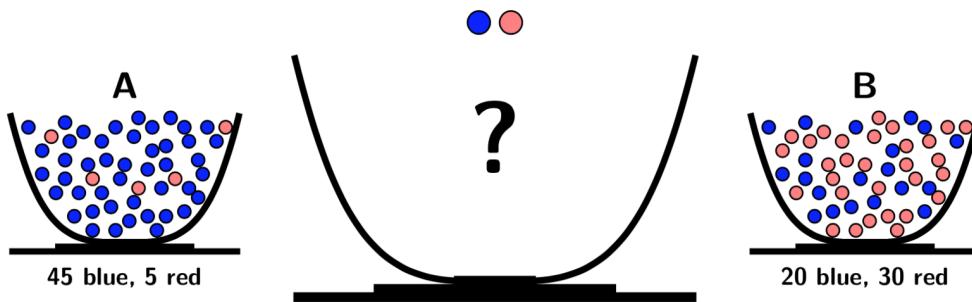
$$\begin{aligned} p(\text{blue}) &= p(\text{blue} | A) * p(A) + p(\text{blue} | B) * p(B) \\ &= p(\text{blue} | A) * p(A) + p(\text{blue} | A^C) * p(A^C) \\ &= 0.65 \end{aligned}$$

# Inverse conditional probability



Someone flips a coin to decide whether to draw a ball from bowl *A* or *B* (each with 50% probability), but *the bowl is hidden from us*.

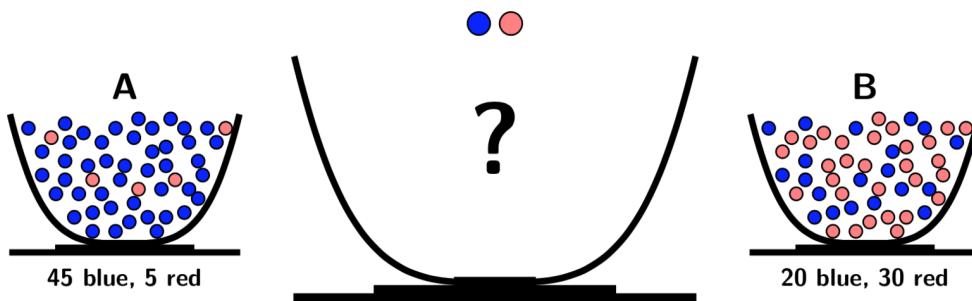
# Inverse conditional probability



Someone flips a coin to decide whether to draw a ball from bowl *A* or *B* (each with 50% probability), but *the bowl is hidden from us*.

- We've drawn a *blue* ball. What's the probability that we drew from *A*?

# Inverse conditional probability



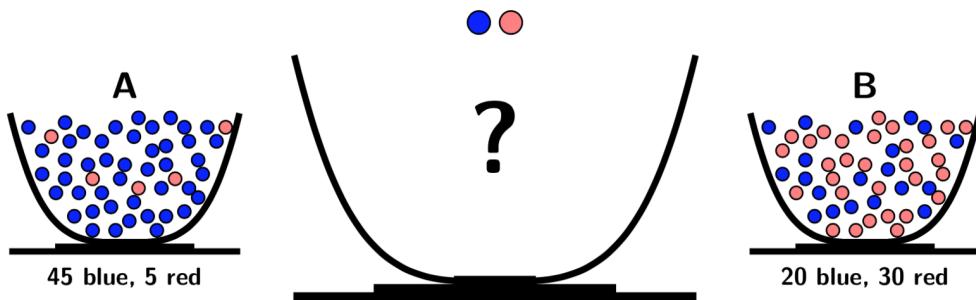
Someone flips a coin to decide whether to draw a ball from bowl  $A$  or  $B$  (each with 50% probability), but *the bowl is hidden from us*.

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"Inverse" conditional probability problem:

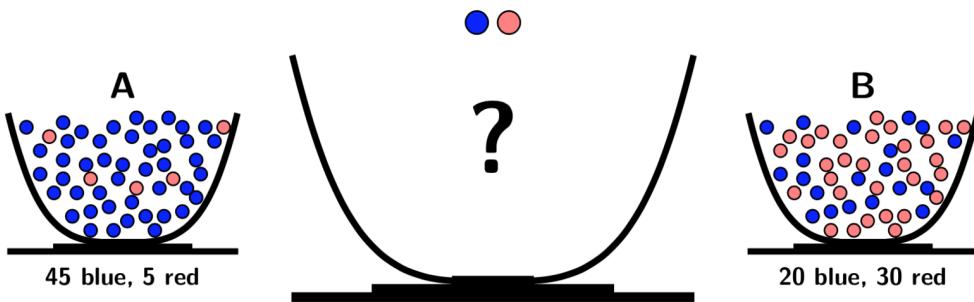
- it's easy to find  $p(\text{blue} \mid A)$ ,
- but how can we *invert* it to find  $p(A \mid \text{blue})$ ?

# Find $p(A \mid \text{blue})$



How do we approach any conditional probability problem?

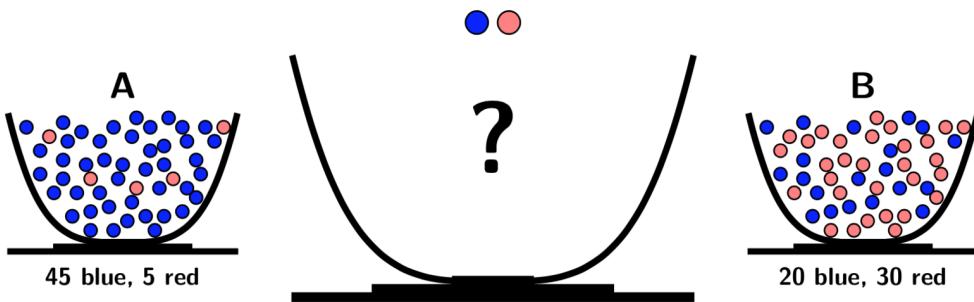
# Find $p(A \mid \text{blue})$



How do we approach any conditional probability problem?

$$p(y \mid x) = \frac{p(y \cap x)}{p(x)}$$

# Find $p(A \mid \text{blue})$

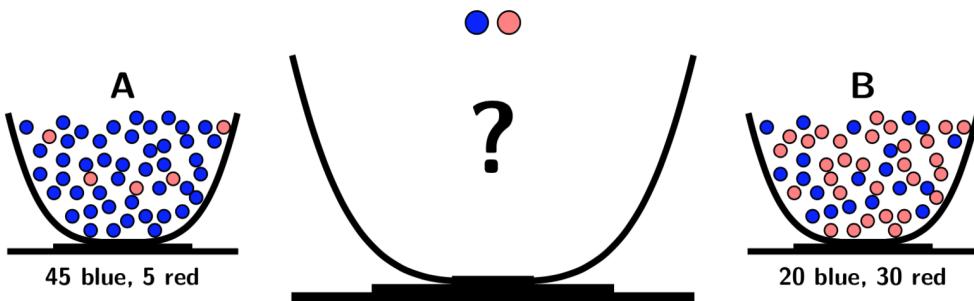


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So what do we need for  $p(A \mid \text{blue})$ ?

# Find $p(A \mid \text{blue})$



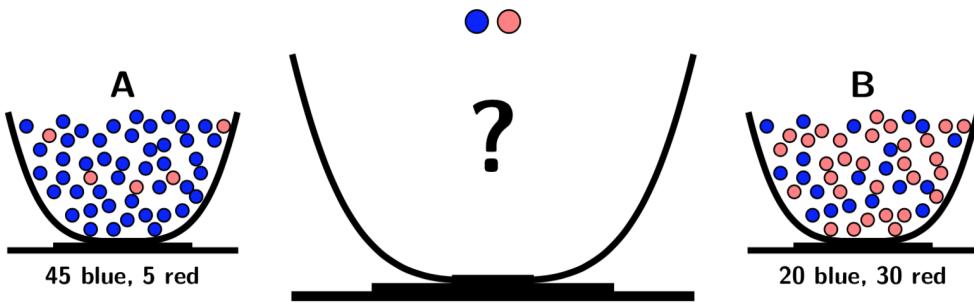
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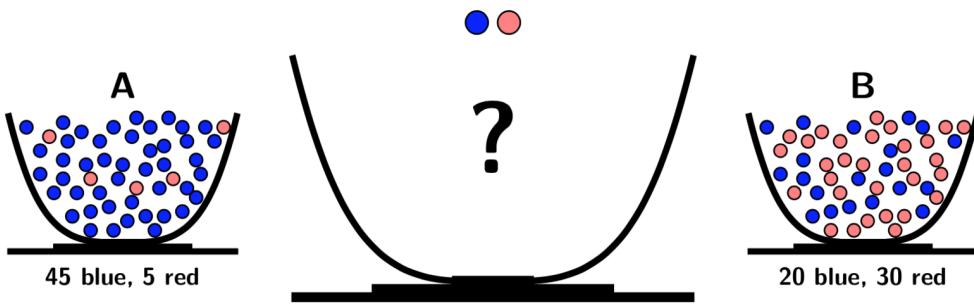
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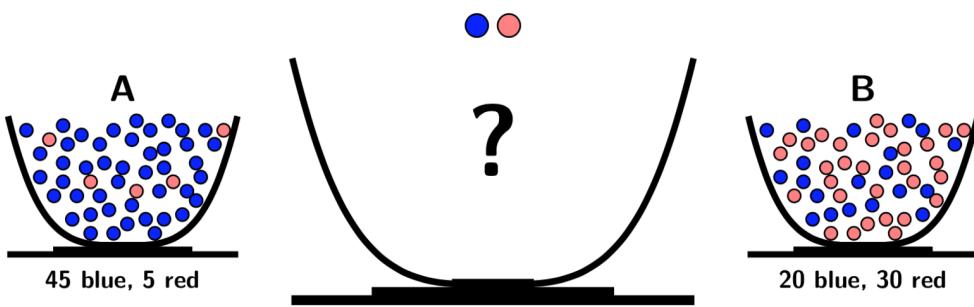
- $p(A \cap \text{blue})$
- $p(\text{blue})$

Find  $p(A \mid \text{blue})$



$p(A \cap \text{blue})?$

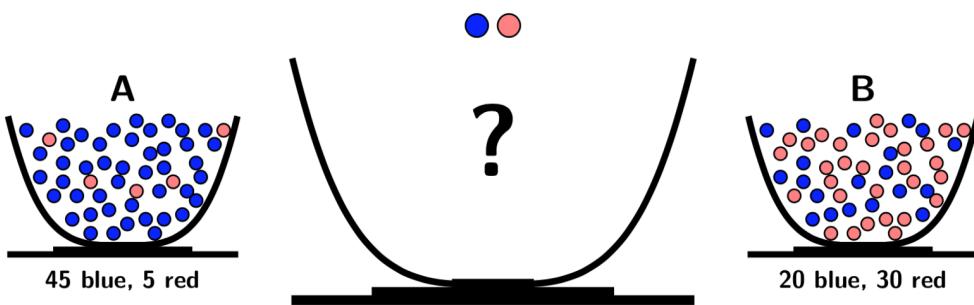
Find  $p(A \mid \text{blue})$



$p(A \cap \text{blue})?$

- $(0.5)(0.9) = 0.45$ , or  $p(\text{blue} \mid A)p(A)$

# Find $p(A \mid \text{blue})$



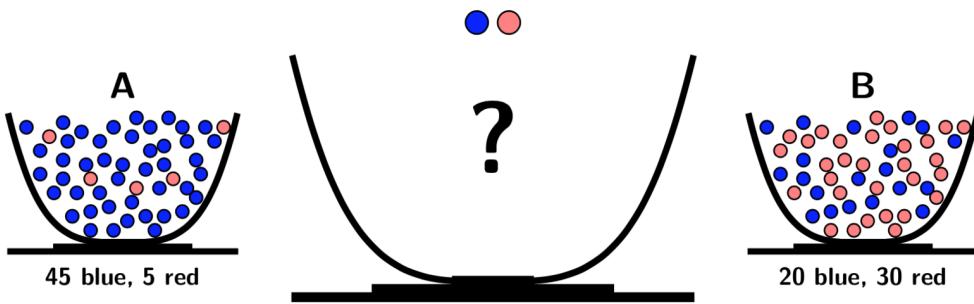
$p(A \cap \text{blue})?$

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$p(\text{blue})?$

- $p(A \cap \text{blue}) + p(B \cap \text{blue})$

# Find $p(A \mid \text{blue})$



$p(A \cap \text{blue})?$

- $(0.5)(0.9) = 0.45$ , or  $p(\text{blue} \mid A)p(A)$

$p(\text{blue})?$

- $p(A \cap \text{blue}) + p(B \cap \text{blue})$
- $(0.5)(0.9) + (0.5)(0.4) = 0.45 + 0.20 = 0.65$

**Find**  $p(A \mid \text{blue})$

$$p(A \mid \text{blue}) = \frac{p(A \cap \text{blue})}{p(\text{blue})}$$

$$p(A \mid \text{blue}) = \frac{p(\text{blue} \mid A)p(A)}{p(\text{blue})}$$

$$p(A \mid \text{blue}) = \frac{0.45}{0.65} \approx 0.69$$

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This is **inverse conditional probability**: how we find  $p(A \mid \text{blue})$  by starting with  $p(\text{blue} \mid A)$ .

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This is **inverse conditional probability**: how we find  $p(A \mid \text{blue})$  by starting with  $p(\text{blue} \mid A)$ .

This is also an example of how **Bayes' Theorem** works. More generally, the theorem is stated as:

$$p(x \mid y) = \frac{p(y \mid x)p(x)}{p(y)}$$

# Bayes' Theorem

Bayes' Theorem describes how to solve the equation by beginning with its inverse

$$p(x | y) = \frac{p(y | x)p(x)}{p(y)}$$

Or, even more generally,

$$p(x | y) = \frac{p(y | x)p(x)}{p(y | x)p(x) + p(y | x^c)p(x^c)}$$

Bayes' Theorem has many applications:

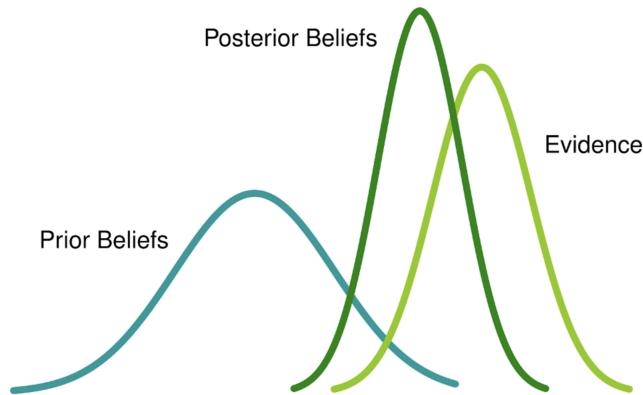
- calculating risks of food allergies or rare diseases
- finding the sources of mechanical errors
- machine learning and prediction; e.g., how likely is a candidate to win an election
- formal modeling (how actors derive their optimal responses)

It is also the basis for Bayesian inference and Bayesian statistics

# Bayesian inference

- We have *prior beliefs*, say, about the probability of a "heads" on a coin. These beliefs are uncertain: there is a range they fall in
- Flip the coin several times, calculate the likelihood of *evidence* given the distribution of prior beliefs
- Calculate revised (posterior) beliefs:

$$\text{Posterior} = \frac{p(\text{evidence} | \text{prior}) \times \text{prior}}{p(\text{evidence})}$$



Plot from Analytics Vidhya

# Two ways to think about statistics

"Frequentism"

- Over a large number of repeated trials, probability is **the fraction of trials** in which an event occurs
- We have some **assumptions about this repeated sampling**, from which statistical properties come
- There exists a **fixed** true parameter (underlying probability of an event), which we estimate
- For different assumed parameter values, we can calculate the probability that our data were created by these values
- If data appear unlikely under certain parameter values, we can reject these parameter values
- Focus is on the probability of the **data**, assuming a fixed parameter

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## "Bayesianism"

- Probability is **our belief** about how likely an event is, given the information we have
- Statistical properties come from the **posterior distribution**
- Parameters are "**random**" (not fixed), only approximated with a distribution, given the model and data
- We have prior notions about plausible parameter values
- We update our prior based on the likelihood of data at different prior values to form posterior beliefs
- Focus is on the probability of the **parameter**, updating priors with data

# Looking ahead

# Quantitative methods courses

If you want to understand statistical work in political science:

- 812, 813, MLE

Formal theory courses:

- 835 (intro to game theory)
- Formal models of international politics (837)

Advanced methods courses include

- Machine learning, Time series, Bayesian analysis, Experimental methods, Multilevel modeling

Courses outside the department:

- Ag econ: applied regression, applied machine learning, applied econometrics more broadly (for a systematic and more advanced econometrics sequence, look at the Econ department)
- Sociology: survey methods
- Statistics: a wide variety of courses on statistical theory and specific methods at different levels

# Quantitative methods pathways

812 is required, 813 and MLE highly recommended

First field: "I want to study *how to study* politics" (develop new statistical estimators, etc.). You still need a substantive interest

Second field: "I want to teach and research about/use advanced methods." Most grad students in the department who take advanced methods training take this path

Minor: 3 courses. Depends on your dissertation focus. See reqs, talk to your advisors

# Advice for quantitative methods courses

Take as many as you feasibly can; don't delay MLE

Even if you a qualitative researcher, the epistemological lessons of large-N analysis are valuable

Pick something you like and get good at it

- Time series, Bayes, formal models, text as data, causal inference, experiments

Do replication projects (in MLE and beyond)

- [Dataverse](#) is a great resource where scholars post data and code for their papers

# Advice for quantitative methods in the *discipline*

Invest in math skills in the beginning (math department has intro to calculus and other classes; self-learning also works)

If you aim for advanced knowledge of methods, classes are necessary but not sufficient:

- Google is your best friend
- Use [Cross Validated](#) for questions on stats and [Stack Overflow](#) for learning R
- Books and online courses help
- Carefully read empirical work with applications of new methods

Take the open science and the "replication crisis" seriously

Learn to organize your workflow. Some recommended resources:

- [The Plain Person's Guide to Plain Text Social Science](#)
- [R for Data Science](#)
- [Intro to LaTeX](#)
- [Git/GitHub for R users](#)

If you might leave academia for data science, consider machine learning and Python