

CSE 312

# Foundations of Computing II

## Lecture 7: Chain Rule and Independence



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Slide Credit: Based on Stefano Tessaro's slides for 312 19au incorporating ideas from Alex Tsun, Rachel Lin, Hunter Schafer & myself 😊

## Announcements

- No concept check today!
- Section tomorrow is **important** with new content that you will need on pset 3, problem 7. Bring your laptops.
- I have to be out of town (and will be largely unreachable) Thursday-Saturday – Aleks will give Friday's lecture!
- Quiz 1 out later next week. Will cover material from the first two problem sets.

## Friday 10/8: Bayes Theorem with Law of Total Probability

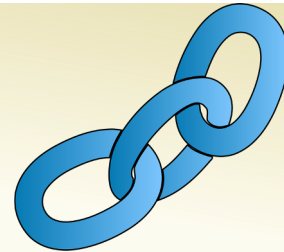
**Bayes Theorem with LTP:** Let  $E_1, E_2, \dots, E_n$  be a partition of the sample space, and  $F$  an event. Then,

$$P(E_1|F) = \frac{P(F|E_1)P(E_1)}{P(F)} = \frac{P(F|E_1)P(E_1)}{\sum_{i=1}^n P(F|E_i)P(E_i)}$$

**Simple Partition:** In particular, if  $E$  is an event with non-zero probability, then

$$P(E|F) = \frac{P(F|E)P(E)}{P(F|E)P(E) + P(F|E^C)P(E^C)}$$

## Monday 10/10: Chain Rule



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

**Theorem. (Chain Rule)** For events  $\mathcal{A}_1, \mathcal{A}_2, \dots, \mathcal{A}_n$ ,

$$\mathbb{P}(\mathcal{A}_1 \cap \dots \cap \mathcal{A}_n) = \mathbb{P}(\mathcal{A}_1) \cdot \mathbb{P}(\mathcal{A}_2|\mathcal{A}_1) \cdot \mathbb{P}(\mathcal{A}_3|\mathcal{A}_1 \cap \mathcal{A}_2)$$

$$\dots \mathbb{P}(\mathcal{A}_n|\mathcal{A}_1 \cap \mathcal{A}_2 \cap \dots \cap \mathcal{A}_{n-1})$$

An easy way to remember: We have  $n$  tasks and we can do them **sequentially**, conditioning on the outcome of previous tasks

## Monday: Independence

**Definition.** Two events  $\mathcal{A}$  and  $\mathcal{B}$  are (statistically) **independent** if

$$\mathbb{P}(\mathcal{A} \cap \mathcal{B}) = \mathbb{P}(\mathcal{A}) \cdot \mathbb{P}(\mathcal{B}).$$

Alternatively,

- If  $\mathbb{P}(\mathcal{A}) \neq 0$ , equivalent to  $\mathbb{P}(\mathcal{B}|\mathcal{A}) = \mathbb{P}(\mathcal{B})$
- If  $\mathbb{P}(\mathcal{B}) \neq 0$ , equivalent to  $\mathbb{P}(\mathcal{A}|\mathcal{B}) = \mathbb{P}(\mathcal{A})$

“The probability that  $\mathcal{B}$  occurs after observing  $\mathcal{A}$ ” -- Posterior  
= “The probability that  $\mathcal{B}$  occurs” -- Prior

## Agenda

- A Sequential Process Defined Using Independence ◀
- Independence As An Assumption
- Sometimes Independence Occurs for Nonobvious Reasons
- Conditional Independence
- Correlation vs Causation
- Information Cascades

## Example – Throwing A Die Repeatedly

Alice and Bob are playing the following game.

A 6-sided die is thrown, and each time it's thrown, regardless of the history, it is equally likely to show any of the six numbers

If it shows 1, 2 → **Alice wins.**

If it shows 3 → **Bob wins.**

Otherwise, **play another round**

What is  $\Pr(\text{Alice wins on } 1^{\text{st}} \text{ round}) =$

...

$\Pr(\text{Alice wins on } i^{\text{th}} \text{ round}) = ?$

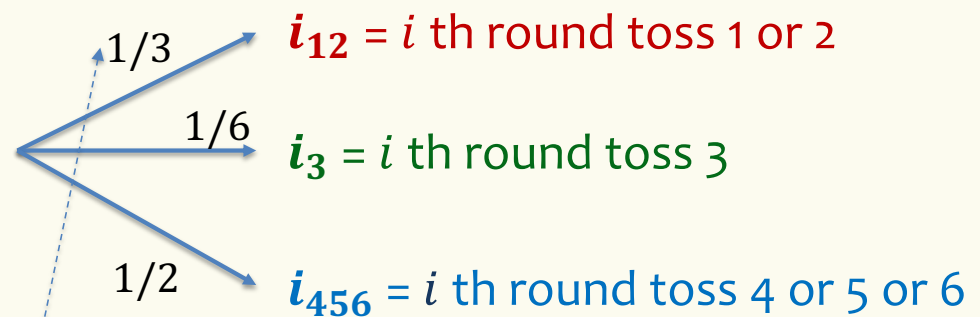
$\Pr(\text{Alice wins}) = ?$

## Sequential Process – defined in terms of independence

A 6-sided die is thrown, and each time it's thrown, regardless of the history, it is equally likely to show any of the six numbers

**Local Rules:** In each round

- If it shows 1,2 → **Alice wins**
- If it shows 3 → **Bob wins**
- Else, play another round

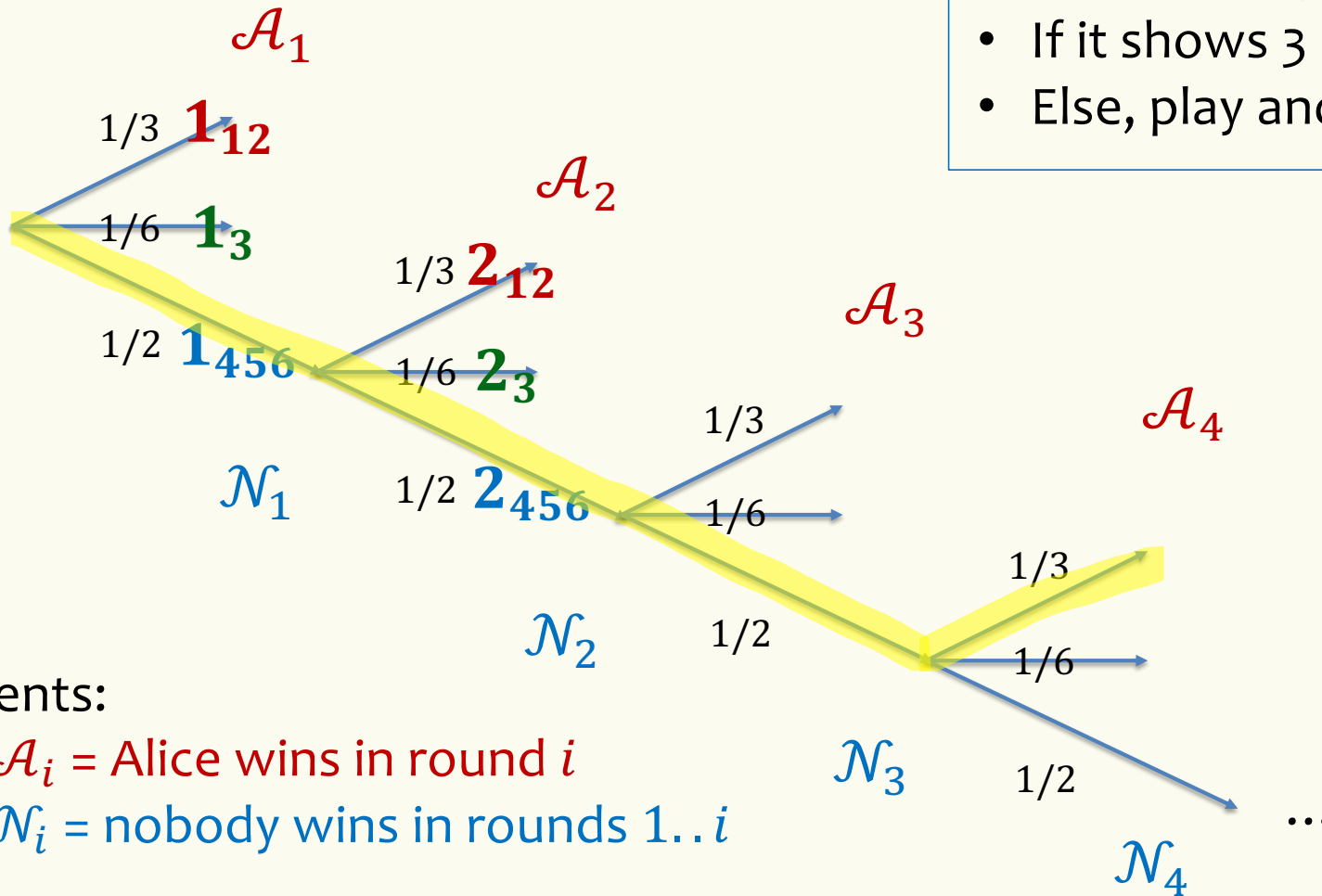


$\Pr(\text{Alice wins on } i\text{-th round} \mid \text{nobody won in rounds } 1..i-1) = \frac{1}{3}$



## Sequential Process – Example

- Local Rules:** In each round
- If it shows 1,2 → **Alice wins**
  - If it shows 3 → **Bob wins**
  - Else, play another round



Events:

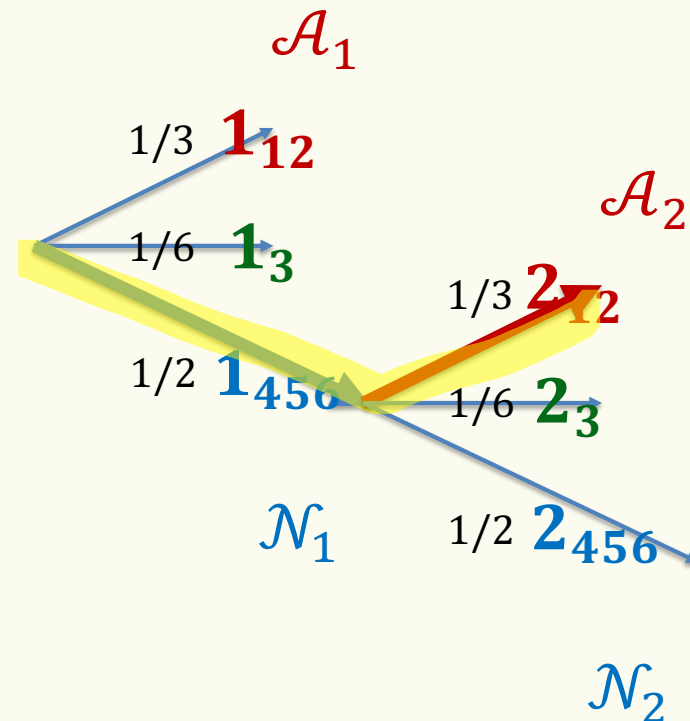
- $\mathcal{A}_i$  = Alice wins in round  $i$
- $\mathcal{N}_i$  = nobody wins in rounds  $1..i$

## Sequential Process – Example

Events:

- $\mathcal{A}_i$  = Alice wins in round  $i$
- $\mathcal{N}_i$  = nobody wins in rounds  $1..i$

$$\mathbb{P}(\mathcal{A}_2) =$$



## Sequential Process – Example

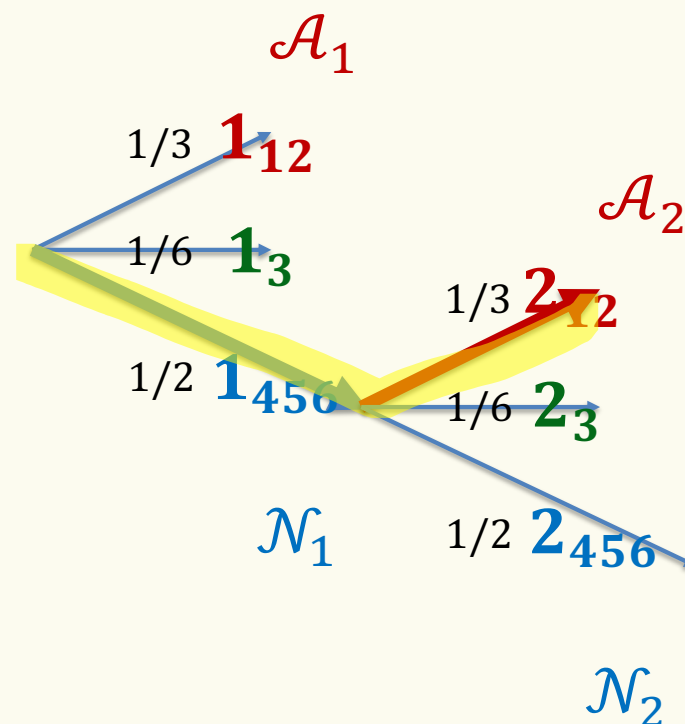
Events:

- $\mathcal{A}_i$  = Alice wins in round  $i$
- $\mathcal{N}_i$  = nobody wins in rounds 1.. $i$

$$\begin{aligned}\mathbb{P}(\mathcal{A}_2) &= \mathbb{P}(\mathcal{N}_1 \cap \mathcal{A}_2) \\ &= \mathbb{P}(\mathcal{N}_1) \times \mathbb{P}(\mathcal{A}_2 | \mathcal{N}_1) \\ &= \frac{1}{2} \times \frac{1}{3} = \frac{1}{6}\end{aligned}$$

The event  $\mathcal{A}_2$  implies  $\mathcal{N}_1$ , and this means that  $\mathcal{A}_2 \cap \mathcal{N}_1 = \mathcal{A}_2$

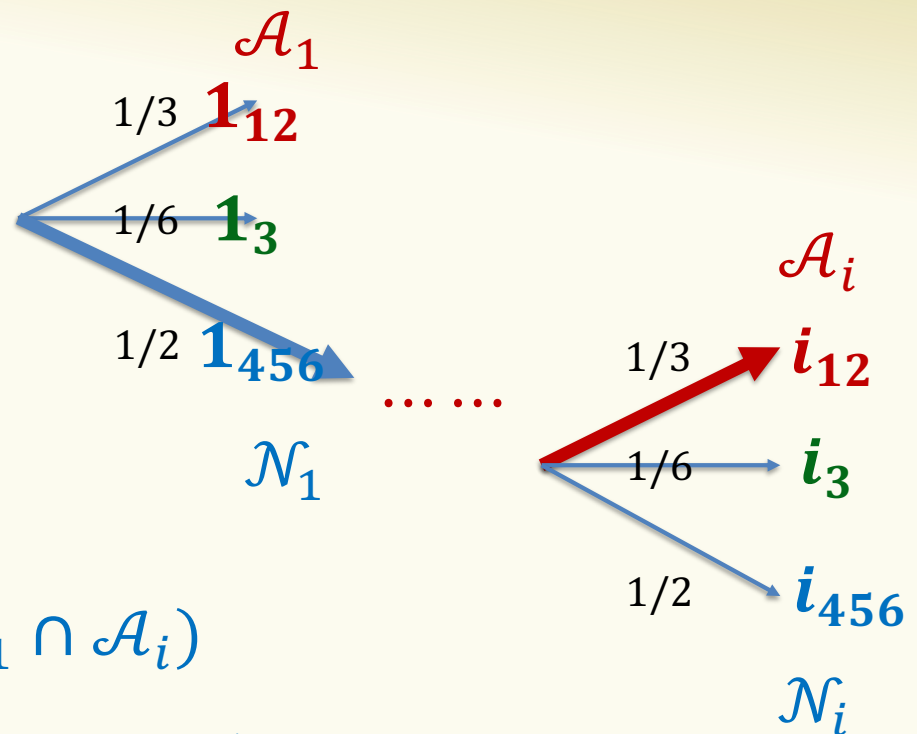
2<sup>nd</sup> roll indep of 1<sup>st</sup> roll



## Sequential Process – Example

Events:

- $\mathcal{A}_i$  = Alice wins in round  $i$
- $\mathcal{N}_i$  = nobody wins in round  $i$



$$\begin{aligned}
 \mathbb{P}(\mathcal{A}_i) &= \mathcal{P}(\mathcal{N}_1 \cap \mathcal{N}_2 \cap \dots \cap \mathcal{N}_{i-1} \cap \mathcal{A}_i) \\
 &= \mathcal{P}(\mathcal{N}_1) \times \mathcal{P}(\mathcal{N}_2 | \mathcal{N}_1) \times \mathcal{P}(\mathcal{N}_3 | \mathcal{N}_1 \cap \mathcal{N}_2) \\
 &\quad \dots \times \mathcal{P}(\mathcal{N}_{i-1} | \mathcal{N}_1 \cap \mathcal{N}_2 \cap \dots \cap \mathcal{N}_{i-1}) \times \mathcal{P}(\mathcal{A}_i | \mathcal{N}_1 \cap \mathcal{N}_2 \cap \dots \cap \mathcal{N}_{i-1}) \\
 &= \left(\frac{1}{2}\right)^{i-1} \times \frac{1}{3}
 \end{aligned}$$

## Sequential Process -- Example

$$\mathcal{A}_i = \text{Alice wins in round } i \quad \mathbb{P}(\mathcal{A}_i) = \left(\frac{1}{2}\right)^{i-1} \times \frac{1}{3}$$

*What is the probability that Alice wins?*

## Sequential Process -- Example

$$\mathcal{A}_i = \text{Alice wins in round } i \quad \mathbb{P}(\mathcal{A}_i) = \left(\frac{1}{2}\right)^{i-1} \times \frac{1}{3}$$

What is the probability that Alice wins?

$$\mathbb{P}(\mathcal{A}_1 \cup \mathcal{A}_2 \cup \dots) = \sum_{i=1}^{\infty} \mathbb{P}(\mathcal{A}_i)$$

All  $\mathcal{A}_i$ 's are disjoint.

$$\sum_{i=1}^{\infty} \left(\frac{1}{2}\right)^{i-1} \times \frac{1}{3} = \frac{1}{3} \times 2 = \frac{2}{3}$$

**Fact.** If  $|x| < 1$ , then  $\sum_{i=0}^{\infty} x^i = \frac{1}{1-x}$ .



## Agenda

- A Sequential Process Defined Using Independence
- Independence As An Assumption ◀
- Sometimes Independence Occurs for Nonobvious Reasons
- Conditional Independence
- Correlation vs Causation
- Information Cascades



## Independence as an assumption

- People often assume it **without justification**.
- Example: A sky diver has two chutes

$A$  : event that the main chute doesn't open

$$\mathbb{P}(A) = 0.02$$

$B$  : event that the backup doesn't open

$$\mathbb{P}(B) = 0.1$$

- What is the chance that at least one opens assuming independence?

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- What is the chance that at least one opens assuming independence?

Assuming independence doesn't justify the assumption! Both chutes could fail because of the same rare event e.g., freezing rain.

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- **Sometimes Independence Occurs for Nonobvious Reasons** ◀
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## Independence – Another Look

**Definition.** Two events  $\mathcal{A}$  and  $\mathcal{B}$  are (statistically) **independent** if

$$\mathbb{P}(\mathcal{A} \cap \mathcal{B}) = \mathbb{P}(\mathcal{A}) \cdot \mathbb{P}(\mathcal{B}).$$

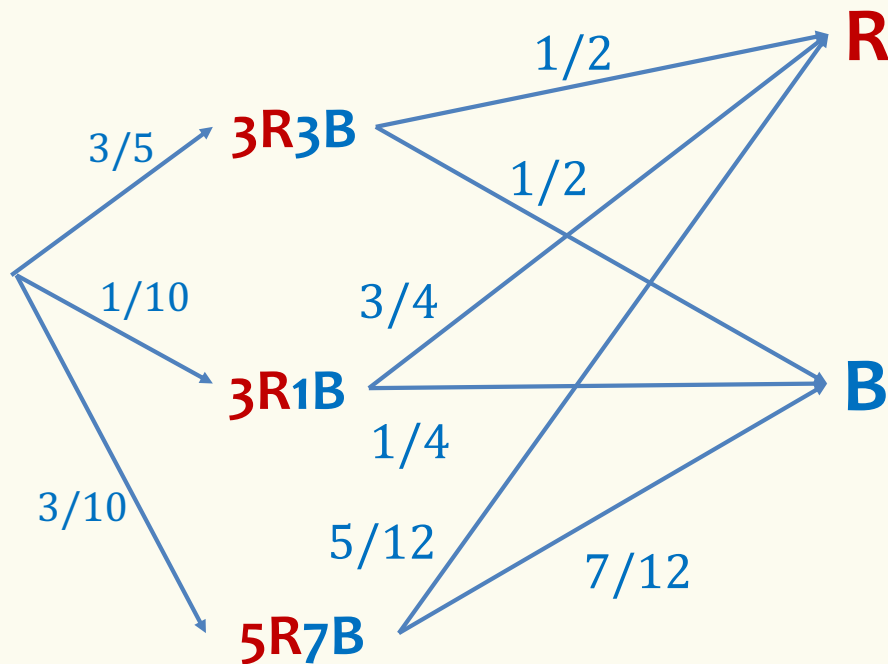
**“Equivalently.”**  $\mathbb{P}(\mathcal{A}|\mathcal{B}) = \mathbb{P}(\mathcal{A}).$

*Events generated independently → their probabilities satisfy independence*

*← Not necessarily*

This can be counterintuitive!

## Sequential Process



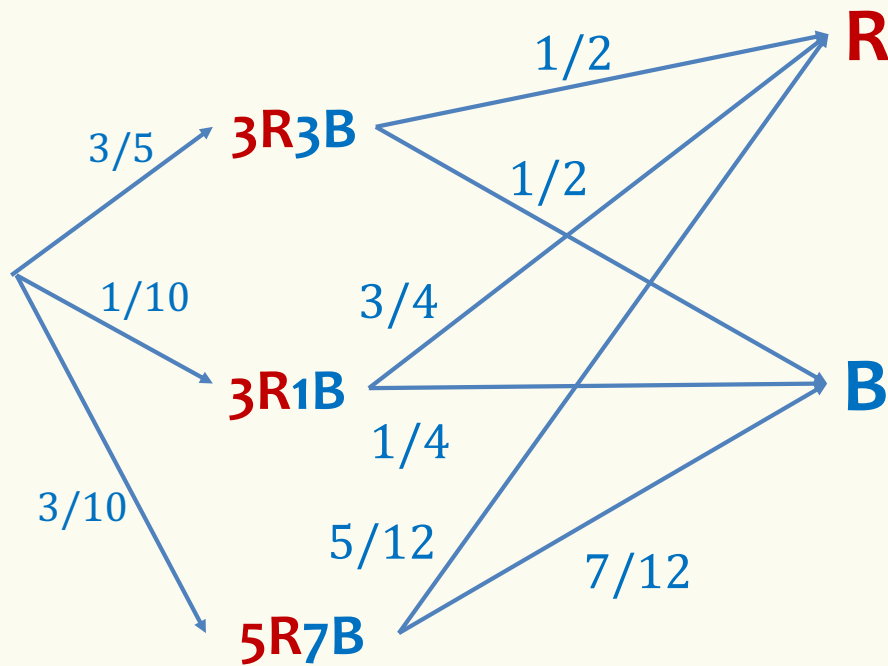
**Setting:** An urn contains:

- 3 **red** and 3 **blue** balls w/ probability  $\frac{3}{5}$
- 3 **red** and 1 **blue** balls w/ probability  $\frac{1}{10}$
- 5 **red** and 7 **blue** balls w/ probability  $\frac{3}{10}$

We draw a ball at random from the urn.

Are  $R$  and  $3R3B$  independent?

## Sequential Process



Are  $R$  and  $3R3B$  independent?

**Setting:** An urn contains:

- 3 **red** and 3 **blue** balls w/ probability  $3/5$
- 3 **red** and 1 **blue** balls w/ probability  $1/10$
- 5 **red** and 7 **blue** balls w/ probability  $3/10$

We draw a ball at random from the urn.

$$\mathbb{P}(R) = \frac{3}{5} \times \frac{1}{2} + \frac{1}{10} \times \frac{3}{4} + \frac{3}{10} \times \frac{5}{12} = \frac{1}{2}$$

$$\mathbb{P}(3R3B) \times \mathbb{P}(R | 3R3B)$$

**Independent!**  $\mathbb{P}(R) = \mathbb{P}(R | 3R3B)$

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- A Sequential Process Defined Using Independence
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- **Conditional Independence** ◀
- Correlation vs Causation
- Information Cascades

## Conditional Independence

**Definition.** Two events  $\mathcal{A}$  and  $\mathcal{B}$  are **independent** conditioned on  $\mathcal{C}$  if  $\mathbb{P}(\mathcal{C}) \neq 0$  and  $\mathbb{P}(\mathcal{A} \cap \mathcal{B} | \mathcal{C}) = \mathbb{P}(\mathcal{A} | \mathcal{C}) \cdot \mathbb{P}(\mathcal{B} | \mathcal{C})$ .



**Plain Independence.** Two events  $\mathcal{A}$  and  $\mathcal{B}$  are **independent** if

$$\mathbb{P}(\mathcal{A} \cap \mathcal{B}) = \mathbb{P}(\mathcal{A}) \cdot \mathbb{P}(\mathcal{B}).$$

Equivalence:

- If  $\mathbb{P}(\mathcal{A}) \neq 0$ , equivalent to  $\mathbb{P}(\mathcal{B} | \mathcal{A}) = \mathbb{P}(\mathcal{B})$
- If  $\mathbb{P}(\mathcal{B}) \neq 0$ , equivalent to  $\mathbb{P}(\mathcal{A} | \mathcal{B}) = \mathbb{P}(\mathcal{A})$



## Conditional Independence

**Definition.** Two events  $\mathcal{A}$  and  $\mathcal{B}$  are **independent** conditioned on  $\mathcal{C}$  if  $\mathbb{P}(\mathcal{C}) \neq 0$  and  $\mathbb{P}(\mathcal{A} \cap \mathcal{B} | \mathcal{C}) = \mathbb{P}(\mathcal{A} | \mathcal{C}) \cdot \mathbb{P}(\mathcal{B} | \mathcal{C})$ .

### Equivalence:

- If  $\mathbb{P}(\mathcal{A} \cap \mathcal{C}) \neq 0$ , equivalent to  $\mathbb{P}(\mathcal{B} | \mathcal{A} \cap \mathcal{C}) = \mathbb{P}(\mathcal{B} | \mathcal{C})$
- If  $\mathbb{P}(\mathcal{B} \cap \mathcal{C}) \neq 0$ , equivalent to  $\mathbb{P}(\mathcal{A} | \mathcal{B} \cap \mathcal{C}) = \mathbb{P}(\mathcal{A} | \mathcal{C})$

## Example – More coin tossing

Suppose there is a coin C1 with  $\Pr(\text{Head}) = 0.3$  and a coin C2 with  $\Pr(\text{Head}) = 0.9$ . We pick one randomly with equal probability and flip that coin twice independently. What is the probability we get all heads?

$$\Pr(HH) = \Pr(HH | C1) \Pr(C1) + \Pr(HH | C2) \Pr(C2)$$

LTP

## Example – More coin tossing

Suppose there is a coin C1 with  $\Pr(\text{Head}) = 0.3$  and a coin C2 with  $\Pr(\text{Head}) = 0.9$ . We pick one randomly with equal probability and flip that coin 2 times independently. What is the probability we get all heads?

$$\Pr(HH) = \Pr(HH | C1) \Pr(C1) + \Pr(HH | C2) \Pr(C2) \quad \text{LTP}$$

$$= \Pr(H | C1)^2 \Pr(C1) + \Pr(H | C2)^2 \Pr(C2) \quad \text{Conditional Independence}$$

$$= 0.3^2 \cdot 0.5 + 0.9^2 \cdot 0.5 = 0.45$$

$$\Pr(H) = \Pr(H | C1) \Pr(C1) + \Pr(H | C2) \Pr(C2) = 0.6$$

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- A Sequential Process Defined Using Independence
- Independence As An Assumption
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- **Correlation vs Causation** ◀
- Information Cascades

## Correlation

- Pick a person at random
- $A$  : event that the person has lung cancer
- $B$  : event that the person is a heavy smoker
  
- Fact:  $\Pr(A|B) = 1.17 \Pr(A)$
  
- Conclusions?

## Correlation

- Pick a person at random
- $A$  : event that the person has lung cancer
- $B$  : event that the person is a heavy smoker
  
- Fact:  $\Pr(A|B) = 1.17 \Pr(A)$
  
- Conclusions?
  - Smoking increases the probability of lung cancer by 17%.
  - Smoking causes lung cancer.

## Correlation

- Pick a person at random
- $A$  : event that the person has lung cancer
- $B$  : event that the person is a heavy smoker
  
- Fact:  $\mathbb{P}(A|B) = 1.17 \cdot \mathbb{P}(A)$
  
- Let's take another look

## Correlation

- Pick a person at random
- $A$  : event that the person has lung cancer
- $B$  : event that the person is a heavy smoker
  
- Fact:  $\mathbb{P}(A|B) = 1.17 \cdot \mathbb{P}(A)$
  
- Conclusions?
  - Lung cancer increases the probability of smoking by 17%.
  - Lung cancer causes smoking.



## Causality vs. Correlation

- Events  $A$  and  $B$  are **positively correlated** if

$$\mathbb{P}(A \cap B) > \mathbb{P}(A) \cdot \mathbb{P}(B)$$

- E.g. smoking and lung cancer.
- But  $A$  and  $B$  being positively correlated does not mean that  $A$  causes  $B$  or  $B$  causes  $A$ .

## Causality vs. Correlation

- Events  $A$  and  $B$  are **positively correlated** if

$$\mathbb{P}(A \cap B) > \mathbb{P}(A) \cdot \mathbb{P}(B)$$

- But  $A$  and  $B$  being positively correlated does not mean that  $A$  causes  $B$  or  $B$  causes  $A$ .

Other examples:

- Tesla owners are more likely to be rich. That does not mean poor people should buy a Tesla to get rich.
- People who go to the opera are more likely to have a good career. That does not mean that going to the opera will improve your career.
- Rabbits eat more carrots and do not wear glasses. Are carrots good for eyesight?

## Proving Causality


- Very difficult!

You have to eliminate external causes of correlation and be able to test the cause/effect relationship (e.g. randomized clinical trials)

Some difficulties:

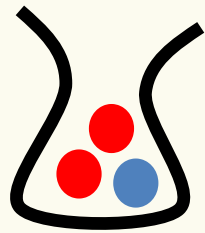
- A and B can be positively correlated because they have a common cause. E.g. being a rabbit.
- If B precedes A, then B is more likely to be the cause (e.g. smoking before lung cancer). However, they could have a common cause that induces both (e.g. studios, take CSE 312)

## Agenda

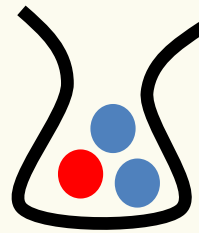
- A Sequential Process Defined Using Independence
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## Guessing game

Experiment: There are 3 balls in this urn. It is a "Red urn" with probability  $\frac{1}{2}$  and a "Blue urn" with probability  $\frac{1}{2}$



Call this a  
"Red urn"



Call this a  
"Blue urn"

## Guessing game

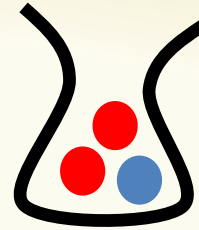
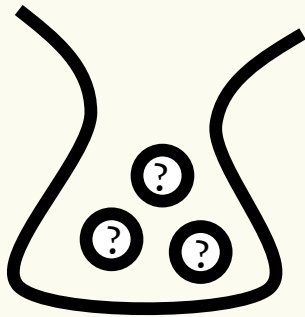
When I call your name:

- You and only you will see a random ball.
- You must then guess if the urn is a Red urn or a Blue urn, and tell the class your guess.

You **do not** tell the class the color of the ball you pulled out

If you guess correctly you will earn one point.

What should the first student do?



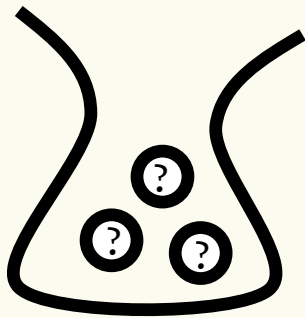
“Red urn”  
Probability  $1/2$

Prior



“Blue urn”  
Probability  $1/2$

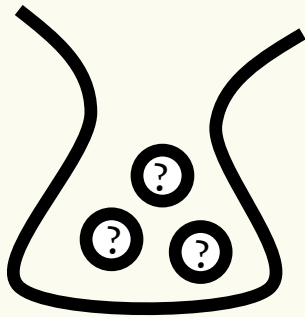
What should the first student do?



Guess that urn is the same color as the ball!



# What should the second student do?



1<sup>st</sup> guess  
was blue.



“Red urn”  
Probability  
 $1/2$

Prior



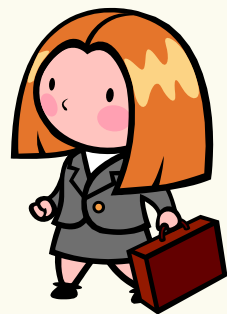
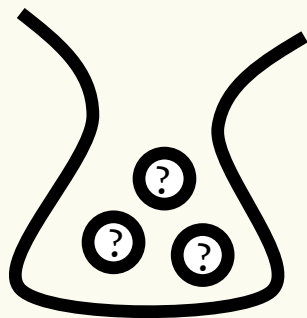
“Blue urn”  
Probability  
 $1/2$

What should the second student do?



Guess that urn is the same color as the ball!

# What should the 3<sup>rd</sup> student do?



1<sup>st</sup> and 2<sup>nd</sup>  
guesses  
were blue.



“Red urn”  
Probability  
 $1/2$

Prior



“Blue urn”  
Probability  
 $1/2$

What should the 3<sup>rd</sup> student do?



Guess that urn is blue, *no matter what she sees!!*

What should the  $n^{\text{th}}$  student do?



If the first two guesses are blue,  
*it is rational for **everyone** to guess blue!*

## Information cascades

- Staring up at the sky
- Choosing a restaurant in a strange city
- Self-reinforcing success of best-selling books.
- Voting for popular candidates

## Information cascades

When

- People make decisions sequentially
- And observe actions of earlier people



**Information Cascade:** People abandon their own information in favor of inferences based on other's actions.



## Information cascades

When

- People make decisions sequentially
- And observe actions of earlier people



**This is rational!!!!**

**Information Cascade:** People abandon their own information in favor of inferences based on other's actions.

## Observations about Information cascades

- Cascades can be wrong.
- Cascades can be based on very little information – if a cascade starts quickly in a large population, most of the private information that is collectively available is not being used.
- Cascades are fragile: Since they can be based on relatively little information, can be easy to stop if some people receive slightly superior information or people reveal more of their private information.

## Monty Hall Problem

*Suppose you're on a game show, and you're given the choice of three doors. Behind one of the doors is a car, behind the other, goats. You pick a door, say number 1, and the host, who knows what's behind the doors, opens another door, say number 3, which has a goat. He says to you, "Do you want to switch to door number 2?" Is it to your advantage to switch your choice of doors?*

### Assumptions

- The player is equally likely to pick each of the three doors.
- After the player picks a door, the host **must** open a different door with a goat behind it and offer the player the choice of staying with the original door he selected or switching to the other unopened door

**Should you switch or stay?**