



# *Probability in Computing*

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## **LECTURE 8**

### **Last time**

- Probability Mass Functions (PMFs)
- Probability Density Functions (PDFs)
- Cumulative Distribution Functions (CDFs)

### **Today**

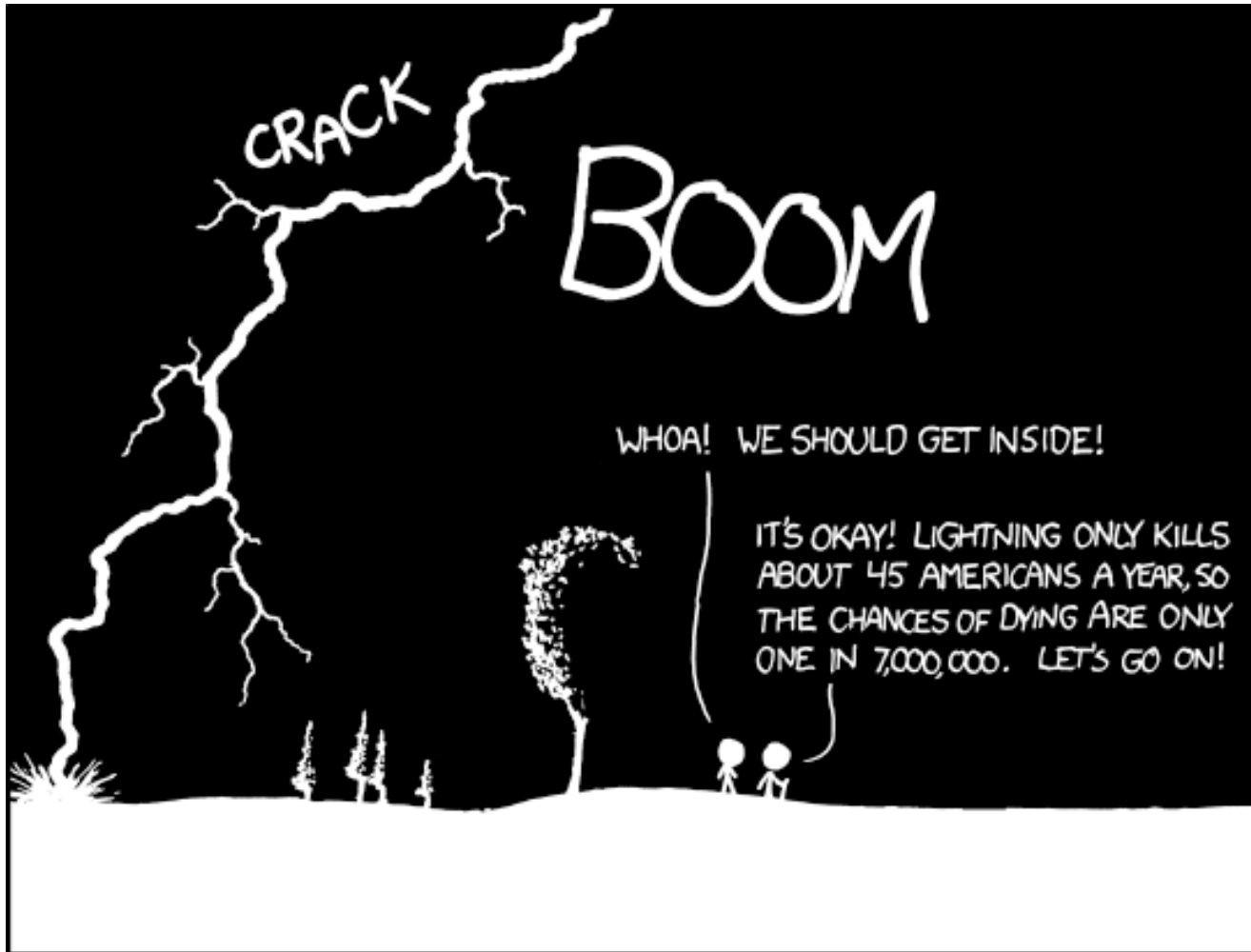
- Conditional Probability

**CS  
237**

### **Reminders**

- HW4 due Thursday

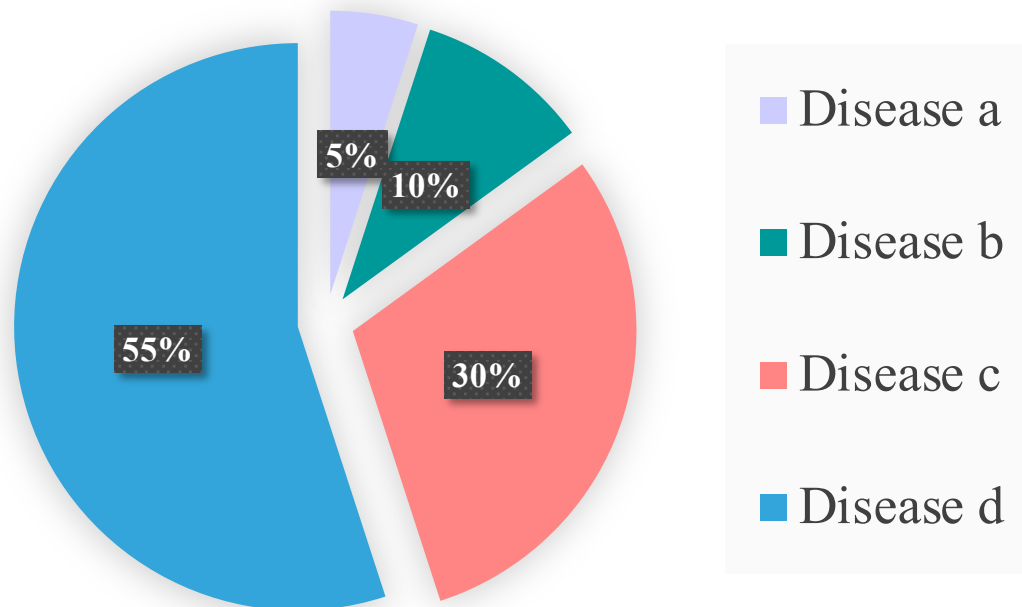
# The annual death rate among people who know this statistic is one in six



<https://xkcd.com/795/>

# Incorporating New Information

- A patient has some unknown disease
- Based on the symptoms, the doctor estimates that the patient has:



- Then a test reveals that the disease is neither **b** nor **d**
- Based on this information, what is the chance of having each disease?

# Incorporating New Information

- A test eliminated diseases **b** and **d**

## Probability

Before the test

After the test

■ Disease a

**0.05**

?

■ Disease b

**0.1**

?

■ Disease c

**0.3**

?

■ Disease d

**0.55**

?

# Incorporating New Information

- A test eliminated diseases **b** and **d**

	Probability	
	Before the test	After the test
■ Disease a	0.05	?
■ Disease b	0.1	0
■ Disease c	0.3	?
■ Disease d	0.55	0

# Incorporating New Information

- A test eliminated diseases **b** and **d**

	Probability	
	Before the test	After the test
■ Disease a	0.05	→ 0.05x
■ Disease b	0.1	→ 0
■ Disease c	0.3	→ 0.3x
■ Disease d	0.55	→ 0

- In the absence of any other information, the new probabilities of a and b should be **proportional** to their original probabilities.

# Incorporating New Information

- A test eliminated diseases **b** and **d**

	Probability	
	Before the test	After the test
■ Disease a	0.05	$\frac{0.05}{0.35}$
■ Disease b	0.1	0
■ Disease c	0.3	$\frac{0.3}{0.35}$
■ Disease d	0.55	0

- Probabilities should add up to 1:  $0.05x + 0.3x = 1$

$$x = \frac{1}{0.35}$$

# Incorporating New Information

- A test eliminated diseases **b** and **d**

	Probability	
	Before the test	After the test
■ Disease a	0.05	$\frac{0.05}{0.35}$
■ Disease b	0.1	0
■ Disease c	0.3	$\frac{0.3}{0.35}$
■ Disease d	0.55	0

- **Note:**  $\Pr(\text{"the disease is neither } b \text{ nor } d\text{"}) = \Pr(\{a, c\}) = 0.35$



# Conditional Probability

- The information provided by the test is that the following event has happened:

$$B = \text{"the disease is neither } b \text{ nor } d\text{"} = \{a, c\}$$

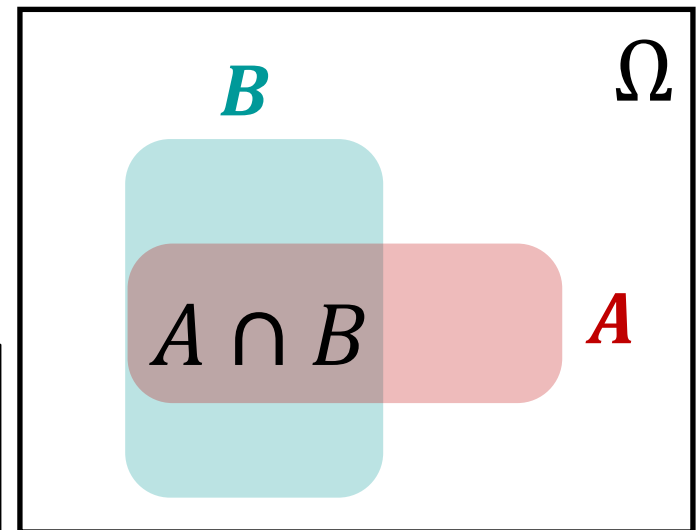
- We define a new probability function that assigns a probability  $\Pr(A \mid B)$  to every event  $A \subseteq \Omega$

$$\Pr(A \mid B) = \frac{\Pr(A \cap B)}{\Pr(B)}$$

Read: "Probability of A given B"

If  $A = \{\omega\}$  for an outcome  $\omega$ ,  
we also write:  $\Pr(\omega \mid B)$

If we specify  $\Pr(\omega \mid B)$  for every outcome  $\omega \in \Omega$ ,  
then  $\Pr(A \mid B)$  is determined for all events  $A$   
(as the sum of probabilities of all outcomes in  $A$ )



# Conditional Probability

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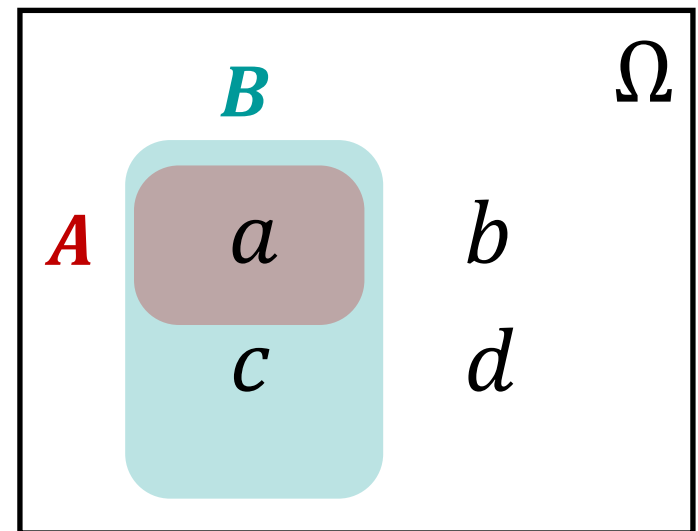
$$B = \text{"the disease is neither } b \text{ nor } d" = \{a, c\}$$

- We define a new probability function that assigns a probability  $\Pr(A | B)$  to every event  $A \subseteq \Omega$

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

$$\Omega = \{a, b, c, d\}$$

$$\Pr(a|B) = \frac{\Pr(a)}{\Pr(B)} = \frac{0.05}{0.35}$$



# Conditional Probability

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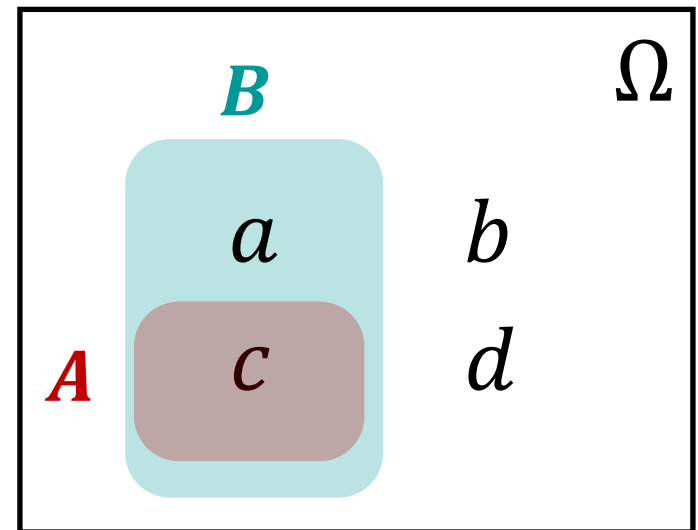
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- We define a new probability function that assigns a probability  $\Pr(A | B)$  to every event  $A \subseteq \Omega$

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

$$\Omega = \{a, b, c, d\}$$

$$\Pr(c|B) = \frac{\Pr(c)}{\Pr(B)} = \frac{0.3}{0.35}$$



# Conditional Probability

- The information provided by the test is that the following event has happened:

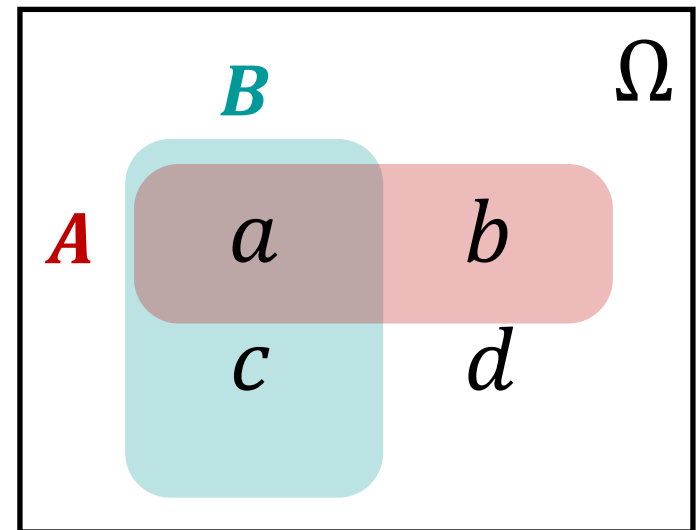
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$$\Omega = \{a, b, c, d\}$$

$$\Pr(\{a, b\} | B) = \frac{\Pr(\{a\})}{\Pr(B)} = \frac{0.05}{0.35}$$



# Conditional Probability

- The information provided by the test is that the following event has happened:

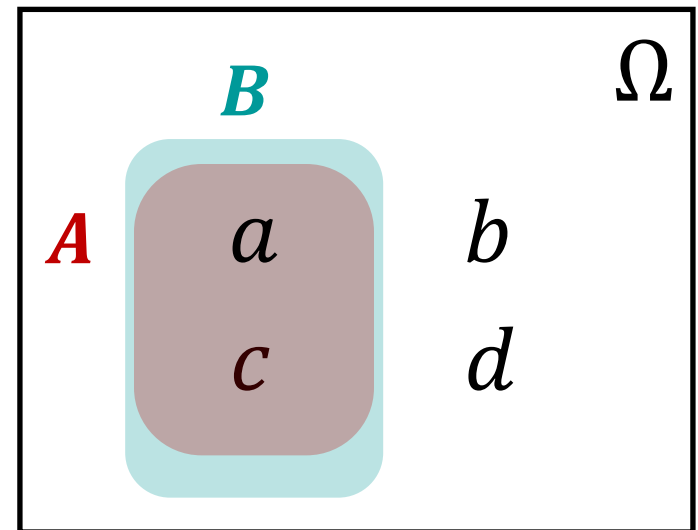
$$B = \text{"the disease is neither } b \text{ nor } d\text{"} = \{a, c\}$$

- We define a new probability function that assigns a probability  $\Pr(A | B)$  to every event  $A \subseteq \Omega$

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

$$\Omega = \{a, b, c, d\}$$

$$\Pr(\{a, c\} | B) = \frac{\Pr(\{a, c\})}{\Pr(B)} = 1$$



## Definition: Conditional Probability

Let  $B$  be an event such that  $\Pr(B) \neq 0$ . For every event  $A$ , we define the **conditional probability of event  $A$  given event  $B$** :

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

**$\Pr(A | B)$  is undefined if  $\Pr(B) = 0$ .**



- Experiment: roll two fair 6-sided dice
- $A = \text{“first die is 6”}$
- $B = \text{“sum of the two rolls is 9”}$

What is  $\Pr(A \mid B)$ ?

A.  $\frac{1}{36}$

B.  $\frac{1}{6}$

C.  $\frac{1}{4}$

D. None of the above



- Experiment: toss a fair coin three times
- $A = \text{“more H than T”}$
- $B = \text{“first toss is H”}$

What is  $\Pr(A \mid B)$ ?

A.  $\frac{3}{8}$

B.  $\frac{1}{2}$

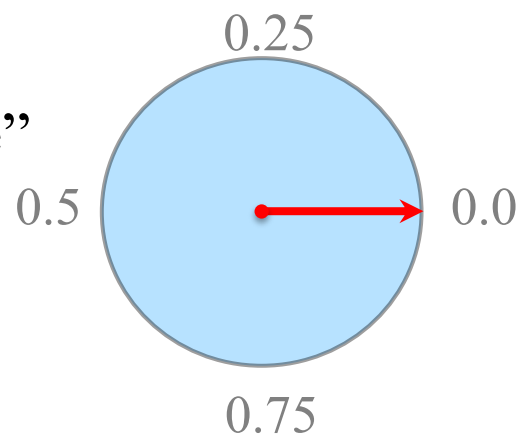
C.  $\frac{3}{4}$

D. None of the above



- Experiment: spin the dial of the spinner
- $A =$  “dial stopped in the upper half of the circle”
- $B =$  “outcome is in  $[1/3, 2/3]$ ”

What is  $\Pr(A \mid B)$ ?

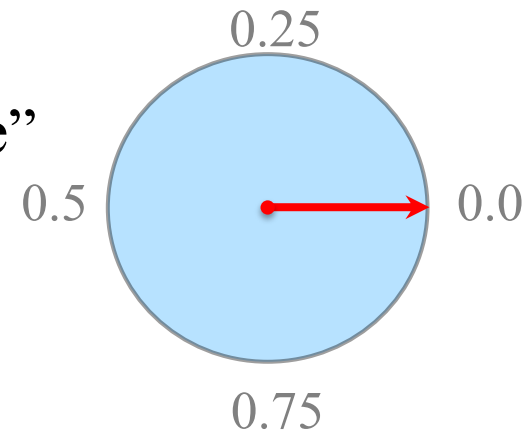


- A.  $\frac{1}{3}$
- B.  $\frac{1}{2}$
- C.  $\frac{3}{4}$
- D. None of the above

# Top Hat question (Join Code: 033357)

- Experiment: spin the dial of the spinner
- A = “dial stopped in the upper half of the circle”
- C = “outcome is in  $[1/6, 1/3]$ ”

What is  $\Pr(A \mid C)$ ?



- A.  $\frac{1}{12}$
- B.  $\frac{1}{6}$
- C. 1
- D. None of the above

Mr. Smith has two children  
(boys and girls are equally likely)

- $A =$  “both children are boys”
- $B =$  “at least one child is a boy”
- $C =$  “older child is a boy”

True or false:  $\Pr(A \mid B) = \Pr(A \mid C)$ ?

- A. True
- B. False



[Martin Gardner]  
[1914-2010]

Conditional probabilities satisfy the axioms of probability

- **Non-negativity:**  $\Pr(A \mid E) \geq 0$  for all events  $A \subseteq \Omega$
- **Additivity:** if  $A$  and  $B$  are *disjoint* events then
$$\Pr(A \cup B \mid E) = \Pr(A \mid E) + \Pr(B \mid E)$$
- **Normalization:**  $\Pr(\Omega \mid E) = 1$
- This applies only to conditional probabilities with the same conditioning event:  $\Pr(\cdot \mid E)$

# Axioms of Probability: Derivation

Conditional probabilities satisfy the axioms of probability

- **Non-negativity:**  $\Pr(A \mid E) \geq 0$  for all events  $A \subseteq \Omega$

**Proof:**

$$\Pr(A \mid E) = \frac{\Pr(A \cap E)}{\Pr(E)}$$

*by definition of  
conditional probability*

- **Normalization:**  $\Pr(\Omega \mid E) = 1$

**Proof:**

$$\Pr(\Omega \mid E) = \frac{\Pr(\Omega \cap E)}{\Pr(E)}$$

*by definition of  
conditional probability*

$$= \frac{\Pr(E)}{\Pr(E)} = 1$$

*since  $E \subseteq \Omega$*

# Axioms of Probability: Derivation

Conditional probabilities satisfy the axioms of probability

– **Additivity**: if  $A$  and  $B$  are *disjoint* events then

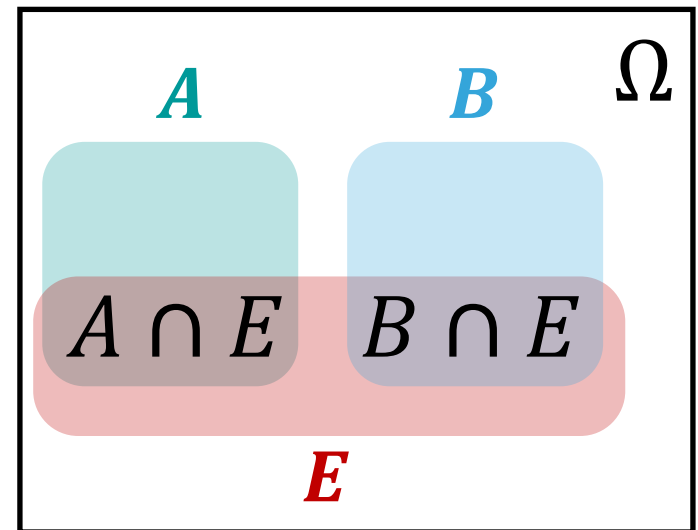
$$\Pr(A \cup B \mid E) = \Pr(A \mid E) + \Pr(B \mid E)$$

**Proof:**

$$\begin{aligned} \Pr(A \cup B \mid E) &= \frac{\Pr((A \cup B) \cap E)}{\Pr(E)} \\ &= \frac{\Pr((A \cap E) \cup (B \cap E))}{\Pr(E)} \\ &= \frac{\Pr(A \cap E) + \Pr(B \cap E)}{\Pr(E)} \\ &= \Pr(A \mid E) + \Pr(B \mid E) \end{aligned}$$

by definition of  
conditional probability

by distributive law



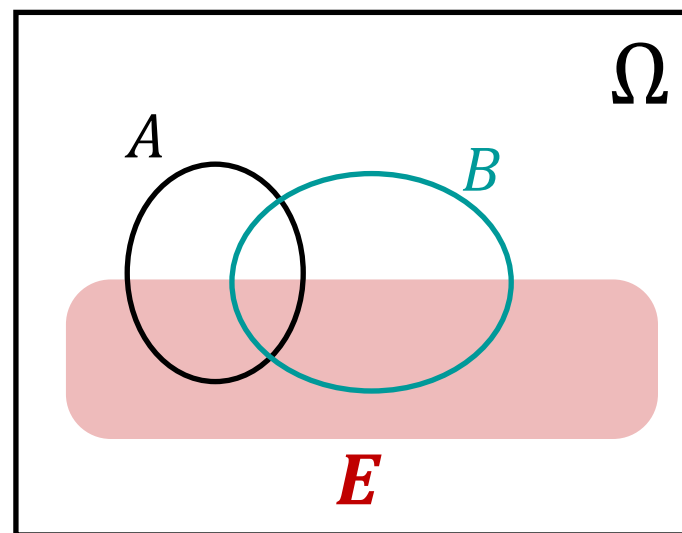
# Conditional Probability Rules

Since the conditional probability function is a valid probability function, all the probability rules remain valid

- Example: **Inclusion-Exclusion Principle**

**Standard:**  $\Pr(A \cup B) = \Pr(A) + \Pr(B) - \Pr(A \cap B)$

**Conditional:**  $\Pr(A \cup B \mid E)$   
 $= \Pr(A \mid E) + \Pr(B \mid E) - \Pr(A \cap B \mid E)$



# Tree Diagrams: Example

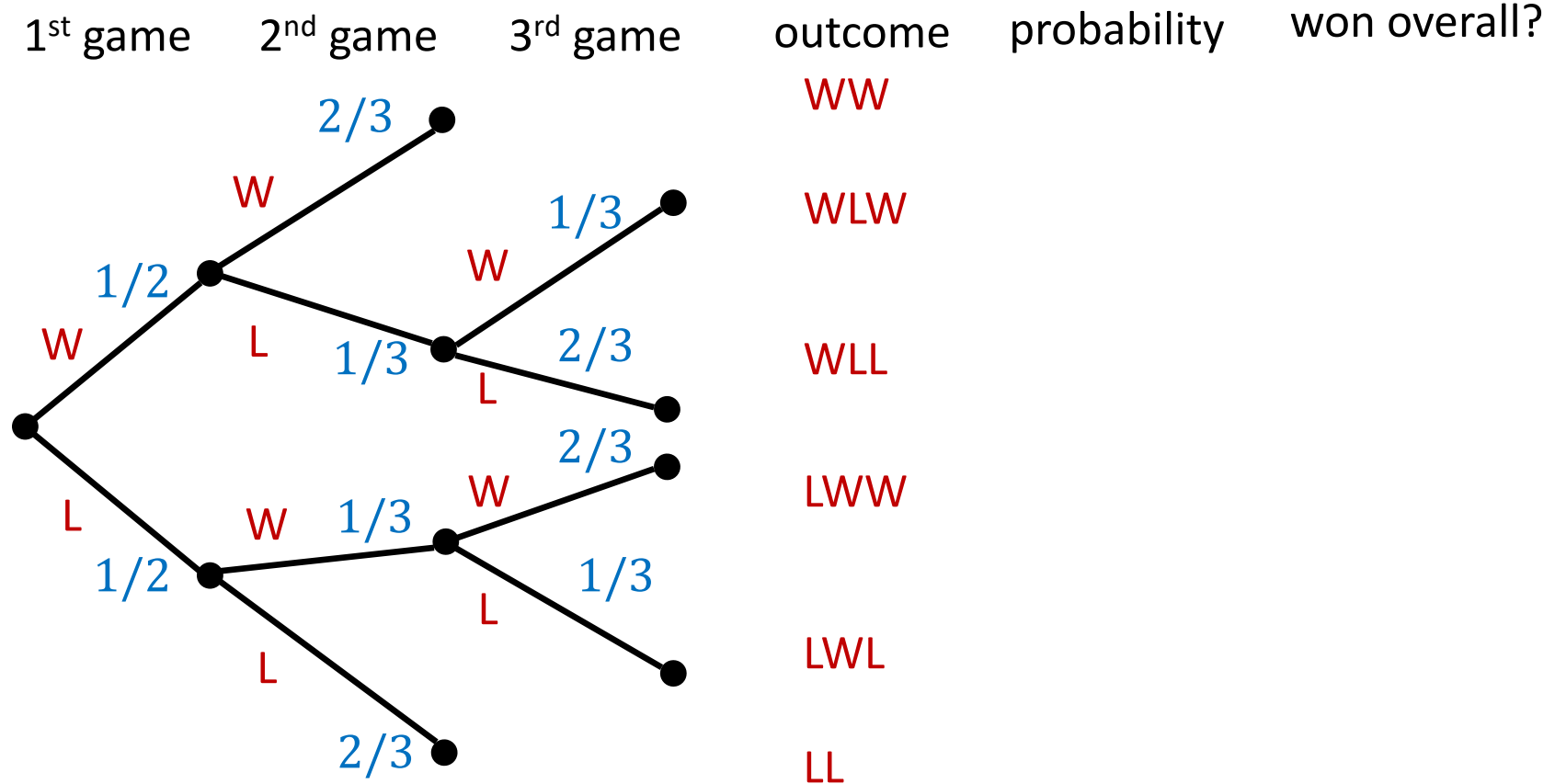
- Best-of-three tournament between BU Terriers and Harvard Huskies
- **First game:** Terriers win with probability  $\frac{1}{2}$
- **Subsequent games:**

Terriers win a game with probability  $\begin{cases} \frac{2}{3} & \text{if they won the previous game} \\ \frac{1}{3} & \text{if they lost the previous game} \end{cases}$

What is the probability that the Terriers win the tournament **given** that they win the first game?



# Tree Diagram: Tournament



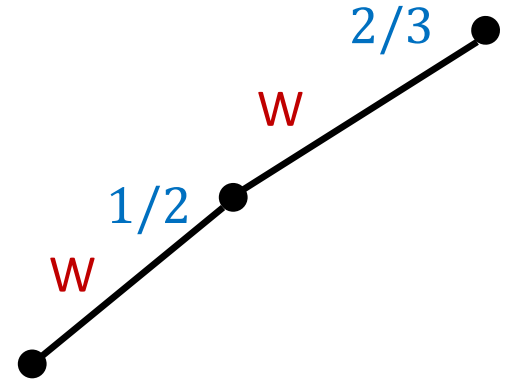
- Event  $T$  = "Terriers win the tournament"
- Event  $E_1$  = "Terriers win the first game"

# Tree Diagram: Justification

- Why do we multiply probabilities along each branch in a tree diagram?

- **Example:**

- Event  $E_1$  = “Terriers win the first game”
- Event  $E_2$  = “Terriers win the second game”



- We calculated  $\Pr(WW) = \Pr(E_1 \cap E_2)$   
 $= \Pr(E_1) \cdot \Pr(E_2 \mid E_1)$
- Product rule follows from the definition of conditional probability:

$$\Pr(E_2 \mid E_1) = \frac{\Pr(E_1 \cap E_2)}{\Pr(E_1)}$$