

DSC 40A

Theoretical Foundations of Data Science I

In This Video

- We'll define the Law of Total Probability and Bayes Theorem.

Getting to Campus

- You conduct a survey:
 - How did you get to campus today? Walk, bike, or drive?
 - Were you late?

	Late	Not Late
Walk	6%	24%
Bike	3%	7%
Drive	36%	24%

Getting to Campus

	Late	Not Late
Walk	6%	24%
Bike	3%	7%
Drive	36%	24%

What is the probability that a randomly selected person is late?

- A. 24%
- B. 30%
- C. 45%
- D. 50%

Getting to Campus

	Late	Not Late
Walk	6%	24%
Bike	3%	7%
Drive	36%	24%

- Since everyone either walks, bikes, or drives,

$$P(\text{Late}) = P(\text{Late AND Walk}) + P(\text{Late AND Bike}) + P(\text{Late AND Drive})$$

- This is called the **Law of Total Probability**.

Getting to Campus

	Late	Not Late
Walk	6%	24%
Bike	3%	7%
Drive	36%	24%

Suppose someone tells you that they walked. What is the probability that they were late?

- A. 6%
- B. 20%
- C. 25%
- D. 45%

Getting to Campus

	Late	Not Late
Walk	6%	24%
Bike	3%	7%
Drive	36%	24%

- Since everyone either walks, bikes, or drives,

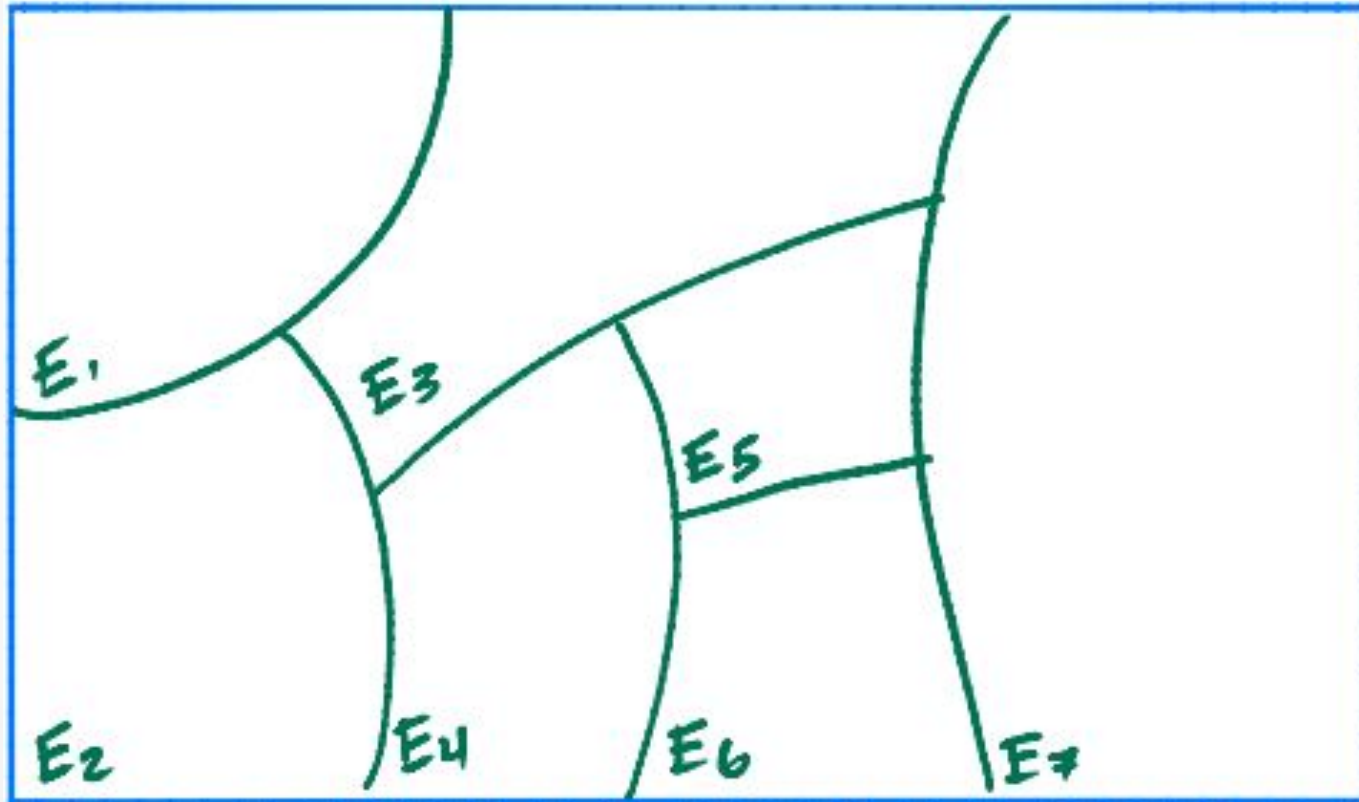
$$P(\text{Late}) = P(\text{Late AND Walk}) + P(\text{Late AND Bike}) + P(\text{Late AND Drive})$$

$$P(\text{Late}) = P(\text{Late}|\text{Walk}) * P(\text{Walk}) + P(\text{Late}|\text{Bike}) * P(\text{Bike}) + P(\text{Late}|\text{Drive}) * P(\text{Drive})$$

Partitions

- A set of events E_1, E_2, \dots, E_k is a **partition** of S if
 - $P(E_i \cap E_j) = 0$ for all i, j
 - $P(E_1) + P(E_2) + \dots + P(E_k) = 1$

Partitions



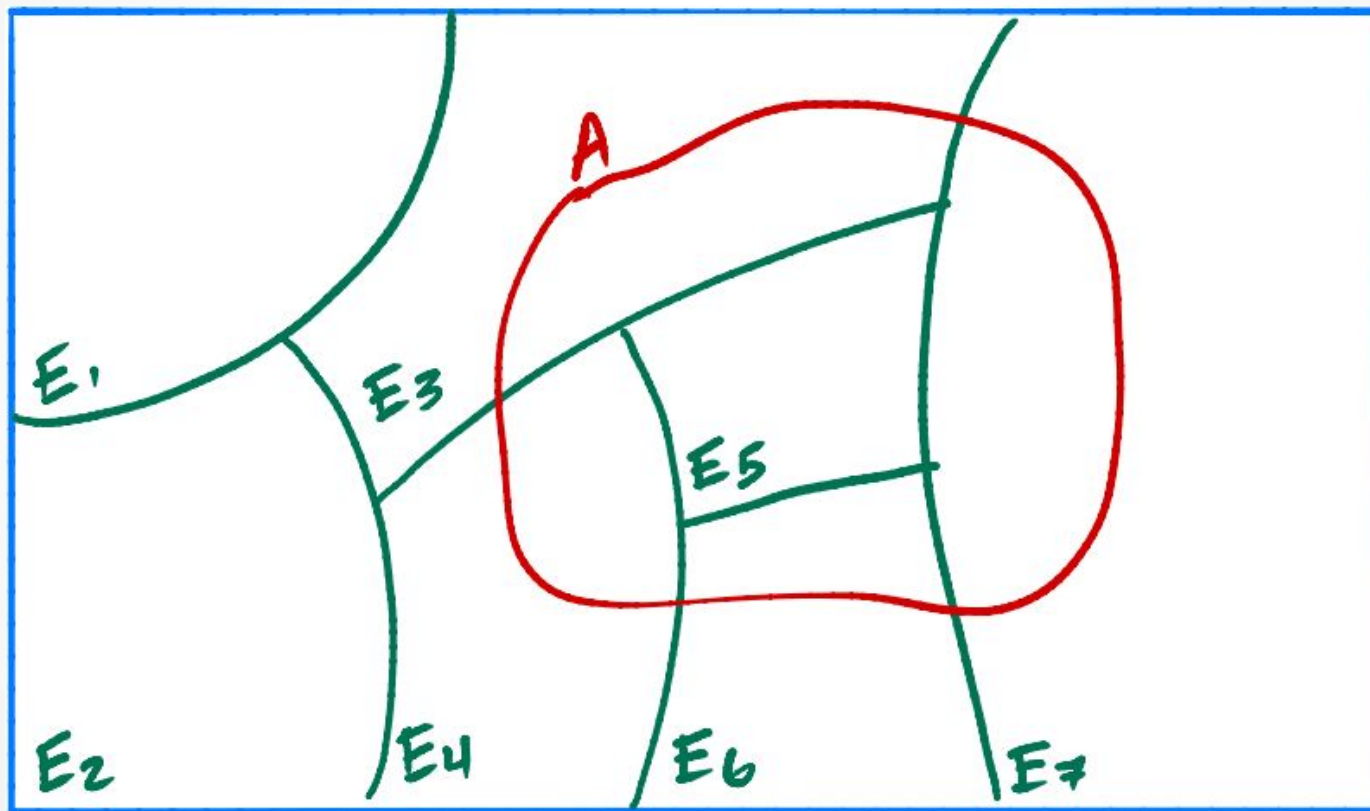
Law of Total Probability

- If A is an event and E_1, E_2, \dots, E_k is a **partition** of S , then

$$P(A) = P(A \cap E_1) + P(A \cap E_2) + \dots + P(A \cap E_k)$$

$$= \sum_{i=1}^k P(A \cap E_i)$$

Partitions



Law of Total Probability

- If A is an event and E_1, E_2, \dots, E_k is a **partition** of S , then

$$\begin{aligned} P(A) &= P(A \cap E_1) + P(A \cap E_2) + \dots + P(A \cap E_k) \\ &= \sum_{i=1}^k P(A \cap E_i) \end{aligned}$$

- Written another way,

$$\begin{aligned} P(A) &= P(A \mid E_1) \cdot P(E_1) + \dots + P(A \mid E_k) \cdot P(E_k) \\ &= \sum_{i=1}^k P(A \mid E_i) \cdot P(E_i) \end{aligned}$$

Getting to Campus

	Late	Not Late
Walk	6%	24%
Bike	3%	7%
Drive	36%	24%

Suppose someone is late. What is the probability that they walked?
Choose the best answer.

- A. Close to 5%
- B. Close to 15%
- C. Close to 30%
- D. Close to 40%

Getting to Campus

- Suppose all you know is
 - $P(\text{Late}) = 45\%$
 - $P(\text{Walk}) = 30\%$
 - $P(\text{Late}|\text{Walk}) = 20\%$
- Can you still find $P(\text{Walk}|\text{Late})$?

Bayes' Theorem

Bayes' Theorem follows from the multiplication rule, or conditional probability.

$$P(A) * P(B|A) = P(A \text{ and } B) = P(B) * P(A|B)$$

Bayes' Theorem:

$$P(B|A) = \frac{P(A|B) * P(B)}{P(A)}$$

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$$P(A) * P(B|A) = P(A \text{ and } B) = P(B) * P(A|B)$$

Bayes' Theorem:

$$\begin{aligned} P(B|A) &= \frac{P(A|B) * P(B)}{P(A)} \\ &= \frac{P(A|B) * P(B)}{P(B) * P(A|B) + P(\overline{B}) * P(A|\overline{B})} \end{aligned}$$

not
B



Bayes' Theorem: Example

$$P(B|A) = \frac{P(A|B) * P(B)}{P(B) * P(A|B) + P(\overline{B}) * P(A|\overline{B})}$$

A manufacturer claims that its drug test will **detect steroid use 95% of the time**.

What the company does not tell you is that 15% of all steroid-free individuals also test positive (the false positive rate). 10% of the Tour de France bike racers use steroids.

Your favorite cyclist just tested positive. What's the probability that he used steroids?

What is your first guess?

- A. Close to 95%
- B. Close to 85%
- C. Close to 40%
- D. Close to 15%

Bayes' Theorem: Example

$$P(B|A) = \frac{P(A|B) * P(B)}{P(B) * P(A|B) + P(\overline{B}) * P(A|\overline{B})}$$

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Now, calculate it and choose the best answer.

- A. Close to 95%
- B. Close to 85%
- C. Close to 40%
- D. Close to 15%

Bayes' Theorem: Example

$$P(B|A) = \frac{P(A|B) * P(B)}{P(B) * P(A|B) + P(\bar{B}) * P(A|\bar{B})}$$

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Solution:

B: used steroids

A: tested positive

Bayes' Theorem: Example

$$P(B|A) = \frac{P(A|B) * P(B)}{P(B) * P(A|B) + P(\bar{B}) * P(A|\bar{B})} = \frac{0.95 * 0.1}{0.1 * 0.95 + 0.9 * 0.15} \approx 0.41$$

A manufacturer claims that its drug test will **detect steroid use 95% of the time**. What the company does not tell you is that 15% of all steroid-free individuals also test positive (the false positive rate). 10% of the Tour de France bike racers use steroids. Your favorite cyclist just tested positive. What's the probability that he used steroids?

Solution:

B: used steroids

A: tested positive

Despite manufacturer's claims, only **41% chance** that cyclist used steroids.

Preview: Bayes' Theorem for Classification

Bayes' Theorem is very useful for classification problems, where we want to predict a class based on some features.

$$P(B|A) = \frac{P(A|B) * P(B)}{P(A)}$$

B = belonging to a certain class
A = having certain features

$$P(\text{class}|\text{features}) = \frac{P(\text{features}|\text{class}) * P(\text{class})}{P(\text{features})}$$

Summary

- When a set of events partitions the sample space, the law of total probability applies.

$$\begin{aligned} P(A) &= P(A \cap E_1) + P(A \cap E_2) + \dots + P(A \cap E_k) \\ &= \sum_{i=1}^k P(A \cap E_i) \end{aligned}$$

- Bayes Theorem says how to express $P(B|A)$ in terms of $P(A|B)$.
- **Next time:** independence and conditional independence