

# Bayes' Rule

## STA 102: Introduction to Biostatistics

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The following material was used by Yue Jiang during a live lecture.

Without the accompanying oral comments, the text is incomplete as a record of the presentation.

## Review: conditional probability

The probability an event will occur when another event has already occurred. The **conditional probability** of event  $A$  given event  $B$  is

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

Examples come up all the time in the real-world:

- ▶ *Given* that a mammogram comes back positive, what is the probability that a woman has breast cancer?
- ▶ *Given* that a 68-year old man has suffered four previous heart attacks, what is the probability he die in the next five years?
- ▶ *Given* that a patient has a mutation in the *CFTR* gene, what is the probability their offspring will have cystic fibrosis?

# Independence and the multiplicative rule

We can rewrite the definition of conditional probability:

$$P(A|B) = \frac{P(A \cap B)}{P(B)} \implies \underbrace{P(A \cap B) = P(A|B) \times P(B)}_{\text{Multiplicative Rule}}$$

What does the multiplicative rule mean in plain English?

Events  $A$  and  $B$  are said to be **independent** when

$$P(A \cap B) = P(A) \times P(B)$$

or equivalently,  $P(A|B) = P(A)$  or  $P(B|A) = P(B)$

## Independent vs. disjoint events

Since for two independent events  $P(A|B) = P(A)$  or  $P(B|A) = P(B)$ , knowing that one event has occurred tells us nothing more about the probability of the other occurring.

For two disjoint events  $A$  and  $B$ , knowing that one has occurred tells us that the other definitely has not occurred:  $P(A \cap B) = 0$

# Conditional probabilities and independence

ORIGINAL RESEARCH

Annals of Internal Medicine

## Coffee Drinking and Mortality in 10 European Countries

A Multinational Cohort Study

What was the probability a randomly selected person in the study...

Coffee drinking	Died		Total
	Yes	No	
None	1039	5438	6477
Med-Low	4440	29712	34152
High	3601	24934	28535
Total	9080	60084	69164

- ▶ ...died?
- ▶ ...died, given they were a non-coffee drinker?

In this study, were dying and coffee drinking independent events?  
How might we check?

# The law of total probability

Suppose we partition  $B$  into mutually disjoint events  $B_1, B_2, \dots, B_k$  that comprise the entirety of \*the entire sample space\*.

The **law of total probability** states that the probability of event  $A$  is

$$\begin{aligned} P(A) &= P(A \cap B_1) + P(A \cap B_2) + \dots + P(A \cap B_k) \\ &= P(A|B_1)P(B_1) + P(A|B_2)P(B_2) + \dots + P(A|B_k)P(B_k) \end{aligned}$$

# The law of total probability in action

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Coffee drinking	Died		Total
	Yes	No	
None	1039	5438	6477
Med-Low	4440	29712	34152
High	3601	24934	28535

What was the probability a randomly selected person died?

## The law of total probability in action

In an introductory statistics course, 50% of students were first years, 30% were sophomores, and 20% were upperclassmen.

80% of the first years didn't get enough sleep, 40% of the sophomores didn't get enough sleep, and 10% of the upperclassmen didn't get enough sleep.

What is the probability that a randomly selected student in this class didn't get enough sleep?

# Bayes' rule

What is the probability that a random person...

Coffee drinking	Died		Total
	Yes	No	
None	1039	5438	6477
Med-Low	4440	29712	34152
High	3601	24934	28535
Total	9080	60084	69164

- ▶ ...was a high coffee drinker, given that they died?
- ▶ ...died, given that they were a high coffee drinker?

These are  $P(High|Died)$  and  $P(Died|High)$ .

Are these two probabilities the same?

# Bayes' rule



(A real portrait of him doesn't seem to exist)

We can use **Bayes' rule** to “reverse” the order of conditioning. By definition:

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

*T. Bayes.*

## Bayes' rule

Using the definition of conditional probability, the law of total probability, and the multiplicative rule, we have

$$\begin{aligned} P(A|B) &= \frac{P(A \cap B)}{P(B)} = \frac{P(B|A)P(A)}{P(B)} \\ &= \frac{P(B|A)P(A)}{P(B|A)P(A) + P(B|A^c)P(A^c)} \end{aligned}$$

If instead  $A$  is partitioned into  $k$  mutually disjoint events that together comprise the entire sample space, then

$$\begin{aligned} P(A|B) &= \frac{P(B|A)P(A)}{P(B)} \\ &= \frac{P(B|A)P(A)}{P(B|A_1)P(A_1) + P(B|A_2)P(A_2) + \cdots + P(B|A_k)P(A_k)} \end{aligned}$$

# Bayes' rule

What is the probability that a random person in the EPIC study...

Coffee drinking	Died		Total
	Yes	No	
None	1039	5438	6477
Med-Low	4440	29712	34152
High	3601	24934	28535
Total	9080	60084	69164

- ▶ ...was a high coffee drinker, given that he died?
- ▶ ...died, given that he was a high coffee drinker?

Verify these results using Bayes' rule

# Looking ahead

