

# Lecture 5: Bayes' rule and independence

## Statistics 251

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# Lecture Outline

Bayes' rule

Independence

# Where are we?

Bayes' rule

Independence

## Recollection on conditional probability

Remember that the conditional probability of  $E$  given  $F$  is

$$\mathbb{P}(E | F) = \frac{\mathbb{P}(E \cap F)}{\mathbb{P}(F)}.$$

Equivalently, we have

$$\mathbb{P}(E \cap F) = \mathbb{P}(F) \cdot \mathbb{P}(E | F).$$

## Law of total probability

Suppose we want to compute  $\mathbb{P}(E)$ . For another event  $F$ , we have

$$\begin{aligned}\mathbb{P}(E) &= \mathbb{P}(E \cap F) + \mathbb{P}(E \cap F^c) \\ &= \mathbb{P}(E | F) \cdot \mathbb{P}(F) + \mathbb{P}(E | F^c) \cdot \mathbb{P}(F^c).\end{aligned}$$

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Example:  $D$  = “have disease”,  $+$  = “test positive”. Suppose  $\mathbb{P}(D) = p$ ,  $\mathbb{P}(+ | D) = 0.9$ , and  $\mathbb{P}(+ | D^c) = 0.1$ . We have

$$\begin{aligned}\mathbb{P}(+) &= \mathbb{P}(+ | D) \cdot \mathbb{P}(D) + \mathbb{P}(+ | D^c) \cdot \mathbb{P}(D^c) \\ &= 0.9p + 0.1(1 - p) = 0.1 + 0.8p.\end{aligned}$$

What we really care about is  $\mathbb{P}(D | +)$ , which is

$$\frac{\mathbb{P}(D \cap +)}{\mathbb{P}(+)} = \frac{\mathbb{P}(+ | D) \cdot \mathbb{P}(D)}{\mathbb{P}(+)} = \frac{0.9p}{0.9p + 0.1(1 - p)}.$$

## Bayes' rule

**Bayes' rule:** By  $\mathbb{P}(B | A) \cdot \mathbb{P}(A) = \mathbb{P}(A \cap B) = \mathbb{P}(A | B) \cdot \mathbb{P}(B)$ , we have

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

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

Interpretation: Start with estimate  $\mathbb{P}(A)$  for  $A$ . After receiving new information, perform a **Bayesian update** to restrict the sample space to  $B$ .

- ▶  $\frac{\mathbb{P}(B|A)}{\mathbb{P}(B)}$  measures how strong the evidence is
- ▶ If  $\mathbb{P}(B | A) = 0$ ,  $A$  and  $B$  are mutually exclusive.
- ▶ We have  $\frac{\mathbb{P}(B|A)}{\mathbb{P}(B)} \leq \frac{1}{\mathbb{P}(A)}$ , with equality if and only if  $\mathbb{P}(A \cap B) = \mathbb{P}(B)$ .



## Bayesian updating

Draw a card at random from a deck. Define the events

$$A = \{\text{card is ace of spades}\}$$

$$B = \{\text{suit of cards is spades}\}.$$

To start, we know that  $\mathbb{P}(A) = \frac{1}{52}$ . If we now know that  $B$  occurred, we may update by

$$\mathbb{P}(A | B) = \frac{\mathbb{P}(B | A)}{\mathbb{P}(B)} \mathbb{P}(A) = \frac{1}{1/4} \cdot \frac{1}{52} = \frac{1}{13}.$$

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## Bayesian updating in the real world...

We can assign probabilities to events which have not yet happened:

$\mathbb{P}(\text{Biden wins election})$

$\mathbb{P}(\text{Cubs win the World Series})$

$\mathbb{P}(\text{stock prices will go up this year})$ .

According to Thomas Bayes:

$\mathbb{P}(A) := \{\text{value of right to get \$1 if event occurs}\}$ .

This creates philosophical questions:

- ▶ Does this “value” have a well-defined price?
- ▶ How is  $\mathbb{P}(A)$  defined when there are no enforceable financial contracts?
- ▶ Can we use this interpretation (and Bayes’ rule) in everyday reasoning?

# Where are we?

Bayes' rule

Independence

## Independent events

Events  $E$  and  $F$  are **independent** if  $\mathbb{P}(E \cap F) = \mathbb{P}(E)\mathbb{P}(F)$ .

- ▶ Equivalent formulation:  $\mathbb{P}(E | F) = \mathbb{P}(E)$
- ▶ Equivalent formulation:  $\mathbb{P}(F | E) = \mathbb{P}(F)$

Toss two coins with sample space  $\{(H, H), (H, T), (T, H), (T, T)\}$ .

- ▶ {first coin H} and {second coin H} are independent
- ▶ {first coin H} and {odd number of H} are independent

Probability of each event is  $\frac{1}{2}$ , and probability of both is  $\frac{1}{4}$ .

## Independence of multiple events

Events  $E_1, \dots, E_n$  are **independent** if for each  $\{i_1, \dots, i_k\} \subset \{1, \dots, n\}$  we have

$$\mathbb{P}(E_{i_1} \cap \dots \cap E_{i_k}) = \mathbb{P}(E_{i_1})\mathbb{P}(E_{i_2}) \cdots \mathbb{P}(E_{i_k}).$$

Implies statements like  $\mathbb{P}(E_1 \cap E_2 \mid E_3 \cap E_4 \cap E_5) = \mathbb{P}(E_1 \cap E_2)$ .

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Does pairwise independence imply independence?

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Does pairwise independence imply independence?

- ▶ No: Consider  $\{\text{first coin H}\}$ ,  $\{\text{second coin H}\}$ ,  $\{\text{odd number of H}\}$



## Independence: examples

Shuffle 4 cards labeled 1, 2, 3, 4. Let

$$E_{i,j} = \{\text{card } i \text{ comes before card } j\}.$$

Is  $E_{1,2}$  independent of  $E_{3,4}$ ?

Is  $E_{1,2}$  independent of  $E_{1,3}$ ?

## Independence: examples

What is  $\mathbb{P}(E_{1,2} \mid E_{1,3})$ ?

What is  $\mathbb{P}(E_{1,7} \mid E_{1,2} \cap E_{1,3} \cap \cdots \cap E_{1,6})$ ?