

# Lecture 6. Independence and Conditional Probability

**YULIA R. GEL**

**CS/SE/STAT 3341 Probability and Statistics  
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## 1 Total Probability

## 2 Bayes Rule

# The "Divide and Conquer Method"

Let  $B_1, B_2, \dots, B_k$  be disjoint and exhaustive events, i.e.  
( $B_i \cap B_j = \emptyset$  for  $i \neq j$ , and  $\cup_{j=1}^k B_j = \Sigma$ ) Then for any event  $A$

$$P(A) = \sum_{j=1}^k P(A \cap B_j) = \sum_{j=1}^k P(A|B_j)P(B_j)$$

This is called **the Law of Total Probability**. It allows us to compute the probability of a complicated event from knowledge of probabilities of simpler events.

# The Chess Example

During the chess tournament, there are three types of opponents for a certain player:

- $P(\text{Type1}) = 0.5$ ,  $P(\text{Win}|\text{Type1}) = 0.3$
- $P(\text{Type2}) = 0.25$ ,  $P(\text{Win}|\text{Type2}) = 0.4$
- $P(\text{Type3}) = 0.25$ ,  $P(\text{Win}|\text{Type3}) = 0.5$

What is probability of player winning?

Solution Using the formula of total probability:

$$P(A) = \sum_{j=1}^3 P(A \cap B_j)$$

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$$\begin{aligned} P(A) &= \sum_{j=1}^3 P(A \cap B_j) = \sum_{j=1}^3 P(A|B_j)P(B_j) \\ &= P(\text{Type1}) \times P(\text{Win}|\text{Type1}) \\ &+ P(\text{Type2}) \times P(\text{Win}|\text{Type2}) \\ &+ P(\text{Type3}) \times P(\text{Win}|\text{Type3}) \end{aligned}$$

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# Thomas Bayes (1702–1761)

Note that  $P(A|B)$  does not need to coincide with  $P(B|A)$ . Now, since  $A \cap B = B \cap A$  and thus  $P(A \cap B) = P(B)P(A|B) = P(B \cap A) = P(A)P(B|A)$ , solving for  $P(B|A)$  gives

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

Using the formula on total probability, we can apply Bayes rule to a general case when we observe  $B_1, B_2, \dots, B_k$  disjoint and exhaustive events (and even to a case when  $k = \infty$ ):

$$P(B_j|A) = \frac{P(B_j)P(A|B_j)}{\sum_{i=1}^k P(B_i)P(A|B_i)}, \quad j = 1, 2, \dots, k.$$

# The Radar Example

Recall that  $B$  is event that the aircraft is flying above and  $A$  is the event that the aircraft is detected by the radar. What is the probability that an aircraft is actually there, given that the radar indicates a detection? Recall  $P(B) = 0.05$ ,  $P(A|B) = 0.99$ ,  $P(A|B^c) = 0.1$ .

Solution Using Bayes rule:

$$P(\text{there is an aircraft} \mid \text{radar detects it}) = P(B|A)$$

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