# Lecture 6 The Bayes' Theorem

**BIO210** Biostatistics

Xi Chen

Fall, 2024

School of Life Sciences
Southern University of Science and Technology

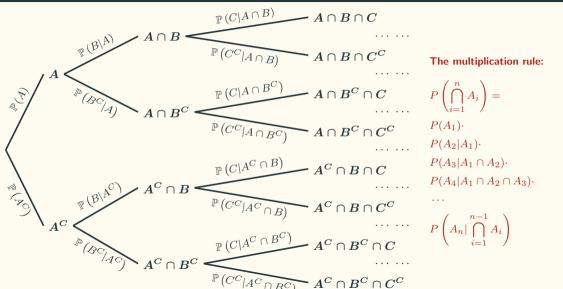


### **Basic components**

Three basic components in conditional probability:

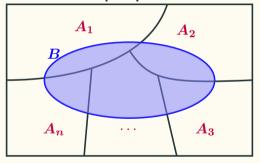
- 1.  $\mathbb{P}(A \cap B)$
- 2.  $\mathbb{P}(B)$
- 3.  $\mathbb{P}(A|B)$

# Generalisation of $\mathbb{P}\left(A\cap B\cap C\cap\cdots ight)$



#### Generalisation of $\mathbb{P}\left(B\right)$

#### Sample space $\Omega$



$$\mathbb{P}(B) = \mathbb{P}[(A_1 \cap B) \cup (A_2 \cap B) \cup \dots \cup (A_n \cap B)]$$

$$= \mathbb{P}(A_1 \cap B) + \mathbb{P}(A_2 \cap B) + \dots + \mathbb{P}(A_n \cap B)$$

$$= \mathbb{P}(A_1) \mathbb{P}(B|A_1) + \dots$$

$$\mathbb{P}(A_2) \mathbb{P}(B|A_2) + \dots$$

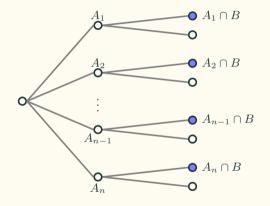
$$\vdots$$

$$\mathbb{P}(A_n) \mathbb{P}(B|A_n)$$

#### The total probability theorem:

$$\mathbb{P}(B) = \sum_{i=1}^{n} \mathbb{P}(B \cap A_i) = \sum_{i=1}^{n} \mathbb{P}(A_i) \mathbb{P}(B|A_i)$$

#### Generalisation of $\mathbb{P}(B)$



$$\mathbb{P}(B) = \mathbb{P}[(A_1 \cap B) \cup (A_2 \cap B) \cup \dots \cup (A_n \cap B)]$$

$$= \mathbb{P}(A_1 \cap B) + \mathbb{P}(A_2 \cap B) + \dots + \mathbb{P}(A_n \cap B)$$

$$= \mathbb{P}(A_1) \mathbb{P}(B|A_1) + \dots$$

$$\mathbb{P}(A_2) \mathbb{P}(B|A_2) + \dots$$

$$\vdots$$

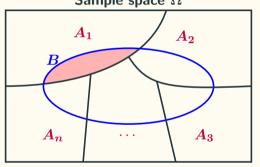
$$\mathbb{P}(A_n) \mathbb{P}(B|A_n)$$

The total probability theorem:

$$\mathbb{P}(B) = \sum_{i=1}^{n} \mathbb{P}(B \cap A_i) = \sum_{i=1}^{n} \mathbb{P}(A_i) \mathbb{P}(B|A_i)$$

# Generalisation of $\mathbb{P}(A|B)$



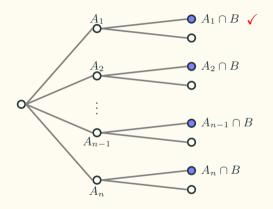


$$\mathbb{P}(A_i|B) = \frac{\mathbb{P}(A_i \cap B)}{\mathbb{P}(B)}$$

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

$$= \frac{\mathbb{P}(A_i) \mathbb{P}(B|A_i)}{\sum_{i=1}^n \mathbb{P}(A_i) \mathbb{P}(B|A_i)}$$

# Generalisation of $\mathbb{P}\left(A|B ight)$

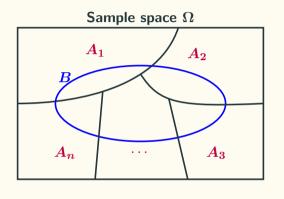


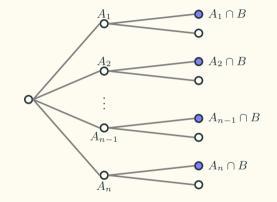
$$\mathbb{P}(A_i|B) = \frac{\mathbb{P}(A_i \cap B)}{\mathbb{P}(B)}$$

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

$$= \frac{\mathbb{P}(A_i) \mathbb{P}(B|A_i)}{\sum_{i=1}^n \mathbb{P}(A_i) \mathbb{P}(B|A_i)}$$

# The Bayes Theorem





$$oldsymbol{A_i} \stackrel{\mathsf{causal\ effect}\ \mathbb{P}\left(B|A_i
ight)}{\mathsf{inference}\ \mathbb{P}\left(A_i|B
ight)} \ oldsymbol{I}$$

#### The Bayes' Theorem

$$\mathbb{P}(A_i|B) = \frac{\mathbb{P}(A_i \cap B)}{\mathbb{P}(B)} = \frac{\mathbb{P}(A_i) \mathbb{P}(B|A_i)}{\sum_{i=1}^n \mathbb{P}(A_i) \mathbb{P}(B|A_i)}$$

 $\mathbb{P}\left(A_{i}
ight)$  : prior probability

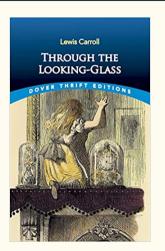
 $\mathbb{P}\left(A_i|B\right)$ : posterior probability

#### Lewis Carroll's Pillow Problems



by Charles Dodgson





#### Lewis Carroll's Pillow Problems

#### Question #5 (8th Sep 1887):

A bag contains one counter, known to be either white or black. A white counter is put in, the bag shaken, and a counter drawn out, which proves to be white. What is now the chance of drawing a white counter?

#### Carroll's Pillow Problem #5

There is a ball inside a non-transparent bag. The colour of the ball is unknown, but it is equally likely to be either blue or red. Now you put a red ball into the bag, shake the bag, and take a ball without looking inside. The ball you have just taken out is red. What is the probability that the colour of the remaining ball that is still inside the bag is red?

# **Pedigree Analysis**

