Lecture 6 The Bayes' Theorem

BIO210 Biostatistics

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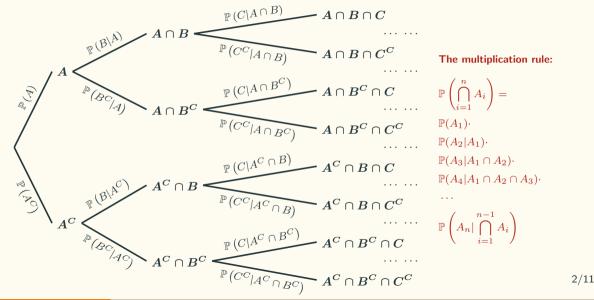


Basic components

Three basic components in conditional probability:

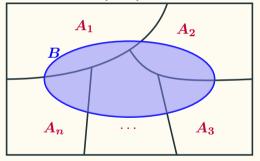
- 1. $\mathbb{P}(A \cap B)$
- 2. $\mathbb{P}(B)$
- 3. $\mathbb{P}(A|B)$

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



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

Sample space Ω



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

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

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

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

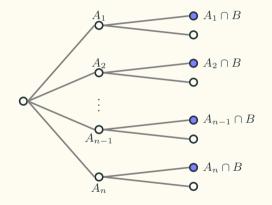
$$\vdots$$

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

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 $\mathbb{P}(B)$



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

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

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

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

$$\vdots$$

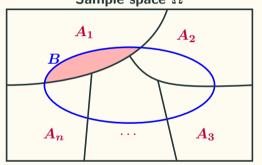
$$\mathbb{P}(A_n) \mathbb{P}(B|A_n)$$

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



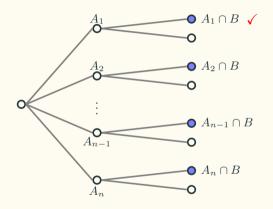


$$\mathbb{P}(A_i|B) = \frac{\mathbb{P}(A_i \cap B)}{\mathbb{P}(B)}$$

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

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

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

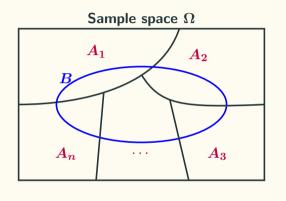


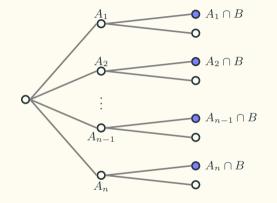
$$\mathbb{P}(A_i|B) = \frac{\mathbb{P}(A_i \cap B)}{\mathbb{P}(B)}$$

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

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

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

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

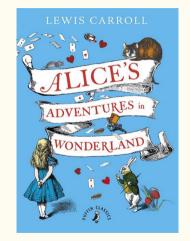
 $\mathbb{P}\left(A_{i}
ight):$ prior probability

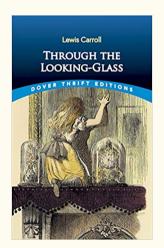
 $\mathbb{P}\left(A_i|B\right)$: posterior probability

Lewis Carroll's Pillow Problems



by Charles Dodgson





Lewis Carroll's Pillow Problems

Question #5 (8th Sep 1887):

2

PILLOW-PROBLEMS.

5. (19, 31)

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?

[8/9/87]

Pedigree Analysis

