

Lecture 6 The Bayes' Theorem

BIO210 Biostatistics

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Spring, 2023

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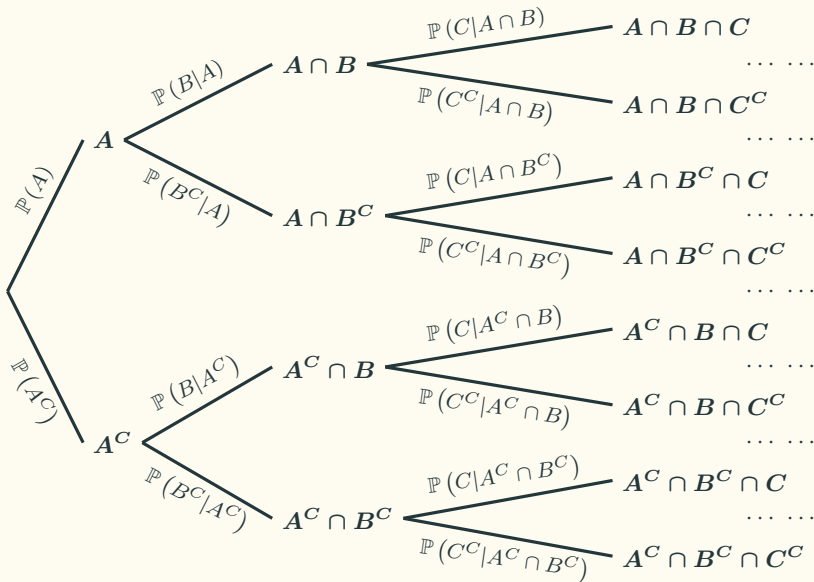


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Three basic components in conditional probability:

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

Generalisation of $\mathbb{P}(A \cap B \cap C \cap \dots)$



The multiplication rule:

$$P\left(\bigcap_{i=1}^n A_i\right) =$$

$$P(A_1) \cdot$$

$$P(A_2|A_1) \cdot$$

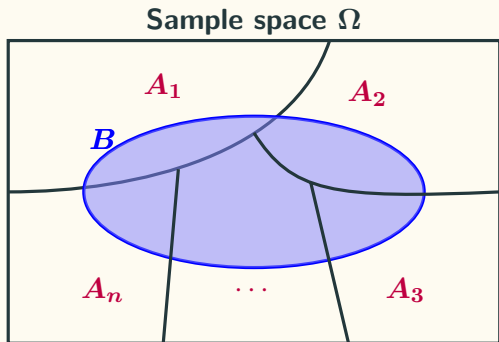
$$P(A_3|A_1 \cap A_2) \cdot$$

$$P(A_4|A_1 \cap A_2 \cap A_3) \cdot$$

\dots

$$P\left(A_n \mid \bigcap_{i=1}^{n-1} A_i\right)$$

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

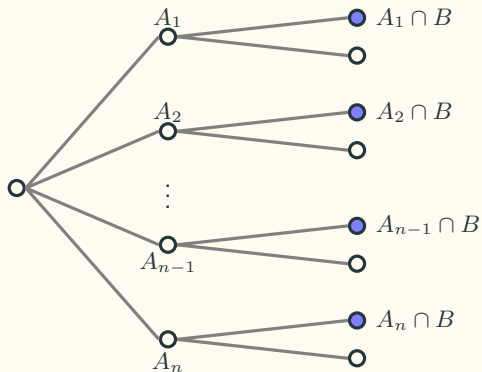


$$\begin{aligned}\mathbb{P}(B) &= \mathbb{P}[(A_1 \cap B) \cup (A_2 \cap B) \cup \dots \cup (A_n \cap B)] \\ &= \mathbb{P}(A_1 \cap B) + \mathbb{P}(A_2 \cap B) + \dots + \mathbb{P}(A_n \cap B) \\ &= \mathbb{P}(A_1) \mathbb{P}(B|A_1) + \\ &\quad \mathbb{P}(A_2) \mathbb{P}(B|A_2) + \\ &\quad \vdots \\ &\quad \mathbb{P}(A_n) \mathbb{P}(B|A_n)\end{aligned}$$

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 $\mathbb{P}(B)$

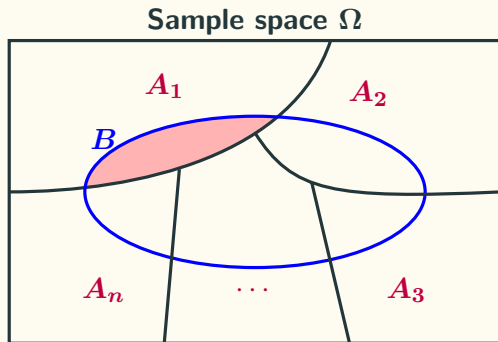


$$\begin{aligned}\mathbb{P}(B) &= \mathbb{P}[(A_1 \cap B) \cup (A_2 \cap B) \cup \dots \cup (A_n \cap B)] \\ &= \mathbb{P}(A_1 \cap B) + \mathbb{P}(A_2 \cap B) + \dots + \mathbb{P}(A_n \cap B) \\ &= \mathbb{P}(A_1) \mathbb{P}(B|A_1) + \\ &\quad \mathbb{P}(A_2) \mathbb{P}(B|A_2) + \\ &\quad \vdots \\ &\quad \mathbb{P}(A_n) \mathbb{P}(B|A_n)\end{aligned}$$

The total probability theorem:

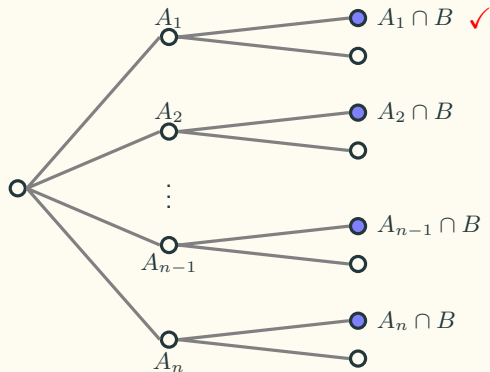
$$\mathbb{P}(B) = \sum_{i=1}^n \mathbb{P}(B \cap A_i) = \sum_{i=1}^n \mathbb{P}(A_i) \mathbb{P}(B|A_i)$$

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



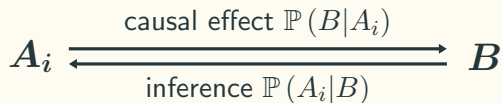
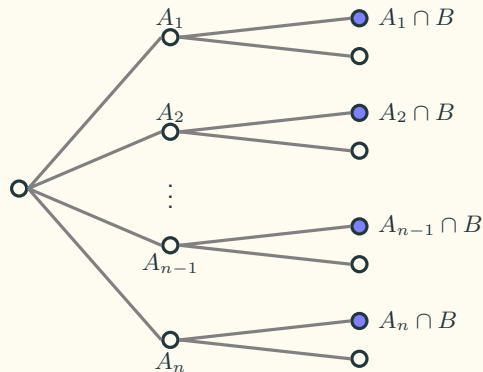
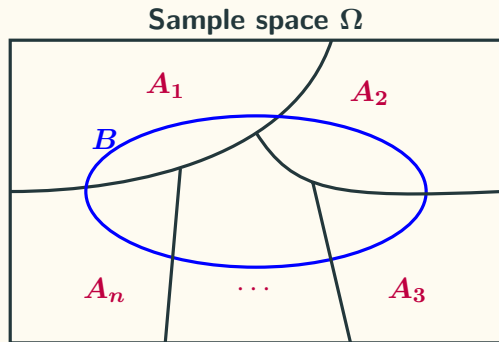
$$\begin{aligned}\mathbb{P}(A_i|B) &= \frac{\mathbb{P}(A_i \cap B)}{\mathbb{P}(B)} \\ &= \frac{\mathbb{P}(A_i) \mathbb{P}(B|A_i)}{\mathbb{P}(B)} \\ &= \frac{\mathbb{P}(A_i) \mathbb{P}(B|A_i)}{\sum_{i=1}^n \mathbb{P}(A_i) \mathbb{P}(B|A_i)}\end{aligned}$$

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



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

The Bayes Theorem



The Bayes' Theorem

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

$\mathbb{P}(A_i)$: **prior probability**

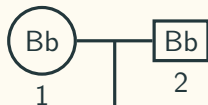
$\mathbb{P}(A_i|B)$: **posterior probability**

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?

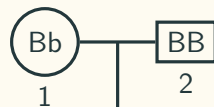
Pedigree Analysis

Generation I



Generation II

Generation I



Generation II



Generation III

