

# Conditional Probability and Independence

"|"  $\triangleq$  "given"

$$P(A|A) = 1 \text{ AND}$$

$$P(B|A) = P(A \cap B | A)$$

↑  
A is now  
sample  
space

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

Thus we get:

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

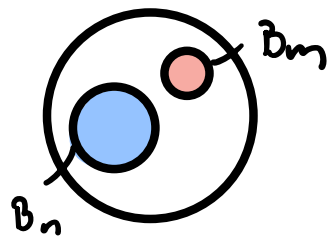
①  $P(B|A) \geq 0$

②  $P(A|A) = 1$

③  $P\left(\bigcup_{n=1}^{\infty} B_n | A\right) = \sum_{n=1}^{\infty} P(B_n | A)$

( $B_1, \dots, B_n$  mutually exclusive)

It follows:  $(B_n \cap A) \cap (B_m \cap A) = \emptyset, n \neq m$



Thus:

$$P\left(\bigcup_{n=1}^{\infty} B_n | A\right) = \frac{P\left(\bigcup_{n=1}^{\infty} (B_n \cap A)\right)}{P(A)}$$

$$= \sum_{n=1}^{\infty} \frac{P(B_n \cap A)}{P(A)}$$

$$= \sum_{n=1}^{\infty} P(B_n | A) \quad \square //$$

## Example

Five card hand, no replacement

↓

all spade hand (B) relative to hypothesis

that four spades are in hand (A) is,

since  $A \cap B = B$

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

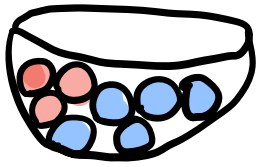
$\{13\}$  all spade hands

$$= \frac{\frac{\binom{15}{5}}{\binom{52}{5}}}{\frac{\binom{13}{4}\binom{39}{1} + \binom{13}{5}}{\binom{52}{5}}} = \frac{\binom{13}{5}}{\binom{13}{4}\binom{39}{1} + \binom{13}{5}} = 0,0441 //$$

We observe

$$P(A \cap B) = P(A)P(B|A)$$

Example 1.4.2 Prob. of red, then blue



$$\Rightarrow P(A) = \frac{3}{8}$$

$$P(B|A) = \frac{5}{7}$$

$$\text{Thus: } P(A \cap B) = \frac{3}{8} \cdot \frac{5}{7} = \frac{15}{56} = 0,2679 //$$

Example 1.4.3

Prob. of third spade appearing on sixth draw:

Let  $A \triangleq \spadesuit \spadesuit$  in first 5 draws

$B \triangleq \spadesuit$  in 6<sup>th</sup> draw

We look for  $P(A \cap B)$ !

$$P(A) = \frac{\binom{13}{2}\binom{39}{3}}{\binom{52}{5}} = 0,2743$$

no more spades

no spades

$$P(B|A) = \frac{13-2}{52-5} = \frac{11}{47} = 0,2340$$

$\nwarrow$  gone  
 $\nearrow$  drawn cards

$$P(B|A) \cdot P(A) = 0,0642 = P(A \cap B)$$

### Example 1.4.4

Drawing 4 cards, P of getting

$\spadesuit \rightarrow \clubsuit \rightarrow \heartsuit \rightarrow \diamondsuit$

$$\Rightarrow \frac{13}{52} \cdot \frac{13}{51} \cdot \frac{13}{50} \cdot \frac{13}{49} = 0,0044 //$$

(Multiplication rule)

Consider  $k$  mut. excl. events (exh. too)

$$A_1, \dots, A_k. \quad P(A_i) > 0$$

Let  $B$  be another event,  $P(B) > 0$

Thus  $B$  occurs with one and only one of the events  $A_1, \dots, A_k$

$$\rightarrow B = B \cap (A_1 \cup \dots \cup A_k)$$

$$= (B \cap A_1) \cup (B \cap A_2) \cup \dots \cup (B \cap A_k)$$

Since  $A_1, \dots, A_k$  mut. excl.

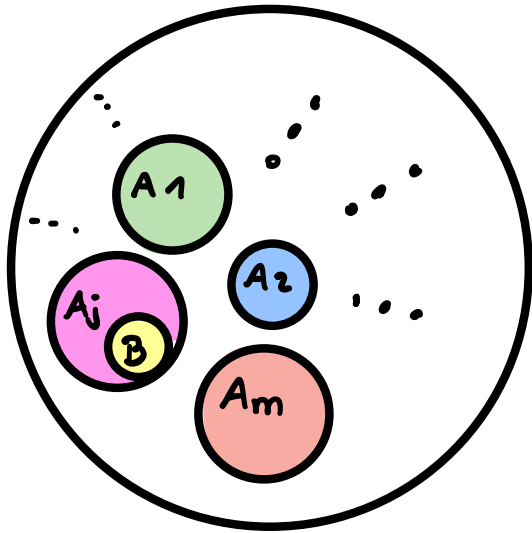
$$\Rightarrow P(B) = P(B \cap A_1) + \dots + P(B \cap A_k)$$

$$+ P(B \cap A_i) = P(A_i) P(B|A_i), i = 1, \dots, k$$

$$\Rightarrow P(B) = P(A_1)P(B|A_1) + \dots + P(A_k)P(B|A_k)$$

$$= \sum_{i=1}^k P(A_i) P(B|A_i) //$$

(Law of total probability)



If  $B \in A_n$  and  $A_m$  or a part of  $B$  even,  $A_n$  and  $A_m$  would **NOT** be mutually exclusive.

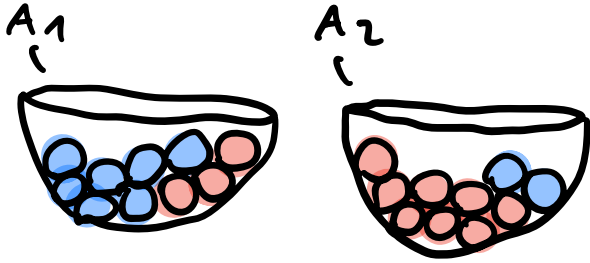
### Theorem 1.4.1 Bayes

$A_1, \dots, A_k$  events,  $P(A_i) > 0$  AND

they form a partition of  $\Omega$ , let  $B$  be any event, then:

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

### Example 1.4.5



Roll die: If  $\therefore$  or  $\ddot{\therefore}$ ,  $A_1$  is selected

to draw from:

$$\rightarrow P(A_1) = \frac{2}{6}$$

$$\rightarrow P(A_2) = \frac{4}{6}$$

Then a chip gets taken ( $B$ )  $\rightarrow$  red chip

$$\rightarrow P(B|A_1) = \frac{3}{10}$$

$$\rightarrow P(B|A_2) = \frac{8}{10}$$

Prob. we drew from bowl  $A_1$ , given  
we drew a red chip, is:

$$P(A_1|B) = \frac{P(A_1)P(B|A_1)}{P(A_1)P(B|A_1) + P(A_2)P(B|A_2)}$$

$$= \frac{\frac{2}{6} \cdot \frac{3}{10}}{\frac{2}{6} \cdot \frac{3}{10} + \frac{4}{6} \cdot \frac{8}{10}} = \frac{3}{19} //$$

$$\text{Thus, also: } P(A_2|B) = \frac{16}{19} //$$

### Example 1.4.6

3 Plants:  $A_1, A_2, A_3$  produce output:  
10%, 50%, 40%.

$A_1$ : 1% of products defective

$A_2$ : 3% defective

$A_3$ : 4% defective

$B$ : selection of random product

$$P(A_1) = 0,1 \quad P(B|A_1) = 0,01$$

$$P(A_2) = 0,5 \quad \text{AND} \quad P(B|A_2) = 0,03$$

$$P(A_3) = 0,4 \quad P(B|A_3) = 0,04$$

Thus,

$$\begin{aligned} P(A_1|B) &= \frac{P(A_1 \cap B)}{P(B)} \\ &= \frac{0,1 \cdot 0,01}{0,1 \cdot 0,01 + 0,5 \cdot 0,03 + 0,4 \cdot 0,04} \\ &= \frac{1}{32} // \end{aligned}$$

### Example 1.4.7

child abused:  $A$

$$P(A) = 0,01$$

not abused:  $N = A^c$

$$P(N) = 0,99$$

Doctor sometimes falsely classifies abuse.

$N_D$ : Abused but D says not abused

$A_D$ : nonabused falsely classified

Suppose:

$$P(N_D|A) = 0,04$$

$$P(A_D|N) = 0,05$$

Thus,

$$P(A_D|A) = 0,96$$

$$P(N_D|N) = 0,95$$

Prob. of random child being classified as abused.

Either  $A \cap A_D$  or  $N \cap A_D$

$$\Rightarrow P(A_D) = P(A_D|A)P(A) + P(A_D|N)P(N)$$

$$= 0,96 \cdot 0,01 + 0,05 \cdot 0,99$$

$$= 0,0591 \approx 6\%$$

(seems quite high)



Prob. of child being abused when doctor classified it:

$$\begin{aligned} P(A|A_D) &= \frac{P(A \cap A_D)}{P(A_D)} \\ &= \frac{0,96 \cdot 0,01}{0,0591} = 0,1624 \\ &\approx 16\% \end{aligned}$$

(seems quite low)

$\Rightarrow$  implies high-error rates  
from doctor

### Independence 1.4.1

$$P(B|A) = P(B)$$

$\rightarrow A, B$  independent

$$\Rightarrow P(A \cap B) = P(A)P(B|A) = P(A)P(B)$$

This implies:

$$\begin{aligned} P(A|B) &= \frac{P(A \cap B)}{P(B)} \\ &= \frac{P(A)P(B)}{P(B)} = P(A) \end{aligned}$$

$$P(A), P(B) > 0$$

$$\Leftrightarrow P(A \cap B) = P(A)P(B) \quad (\text{Independent})$$

$$B = (A^c \cap B) \cup (A \cap B)$$

$$\begin{aligned} \Rightarrow P(A^c \cap B) &= P(B) - P(A \cap B) \\ &= P(B) - P(A)P(B) \\ &= P(B)(1 - P(A)) \\ &= P(B)P(A^c) \end{aligned}$$

$\rightarrow A^c, B$  also independent

### Example 1.4.8

red die, white die  $\Leftrightarrow$  die colour doesn't influence  $P(X)$   
 $P(A) = \frac{1}{6}, P(B) = \frac{1}{6}$   
 red = ::, white = ::

$$P((4,3)) = \frac{1}{6} \cdot \frac{1}{6} = \frac{1}{36}$$

Suppose 3 events now,  $A_1, A_2, A_3$

mutually independent

$\Leftrightarrow$

$$P(A_1 \cap A_3) = P(A_1)P(A_3)$$

$$P(A_1 \cap A_2) = P(A_1)P(A_2)$$

$$P(A_2 \cap A_3) = P(A_2)P(A_3)$$

$$P(A_1 \cap A_2 \cap A_3) = P(A_1)P(A_2)P(A_3)$$

} pairwise independent

$A_1, \dots, A_n$  mut. ind.

$\Leftrightarrow$

$\forall$  collections of  $k$  events,  $2 \leq k \leq n$

$\wedge \forall$  permutations  $d_1, \dots, d_k$ :

$$P(A_{d_1} \cap \dots \cap A_{d_k}) = P(A_{d_1}) \cdot \dots \cdot P(A_{d_k})$$

### Example 1.4.3

Pairwise ind. does NOT imply mutual ind.

We spin a spinner with numbers 1, 2, 3, 4 twice.

$A_1$ : sum is 5

$A_2$ : 1<sup>st</sup> number is 1

$A_3$ : 2<sup>nd</sup> number is 4

$$P(A_i) = \frac{1}{4}, i = 1, 2, 3 \quad \wedge$$

for  $i \neq j$

$$P(A_i \cap A_j) = \frac{1}{16}$$

$\rightarrow$  pairwise independence BUT

$P(A_1 \cap A_2 \cap A_3)$  is event, that (1, 4) is spun

$$\text{which is } \frac{1}{16} \neq \frac{1}{64} = P(A_1)P(A_2)P(A_3)$$

$\rightarrow$  NOT mutually independent

### Example 1.4.10

coin flipped

$A_i = H$  on  $i$ 'th toss

$A_i^c = T$  on  $i$ 'th toss

$$P(A_i) = P(A_i^c) = \frac{1}{2}$$

Thus:

$$\begin{aligned} HNTTH &= P(A_1 \cap A_2 \cap A_3^c \cap A_4) \\ &= P(A_1)P(A_2)P(A_3^c)P(A_4) \\ &= \left(\frac{1}{2}\right)^4 = \frac{1}{16} \end{aligned}$$

Prob. of observing First head on 3<sup>rd</sup> Flip:

$$\begin{aligned} &P(A_1^c \cap A_2^c \cap A_3) \\ &= P(A_1^c)P(A_2^c)P(A_3) \\ &= \left(\frac{1}{2}\right)^3 = \frac{1}{8} \end{aligned}$$

Prob. of  $\geq 1$  H on four flips:

$$\begin{aligned} &P(A_1 \cup A_2 \cup A_3 \cup A_4) \\ &= 1 - P((A_1 \cup A_2 \cup A_3 \cup A_4)^c) \quad \begin{array}{l} \text{de Morgan} \\ (A \cup B)^c = A^c \cap B^c \end{array} \\ &= 1 - P(A_1^c \cap A_2^c \cap A_3^c \cap A_4^c) \\ &= 1 - \left(\frac{1}{2}\right)^4 = \frac{15}{16} \end{aligned}$$

### Example 1.4.11

Computer system built, s.t. if Component  $K_1$  fails,  $K_2$  is used.  $K_2$  fails  $\rightarrow K_3$  used...

$$\left. \begin{array}{l} P(K_1 \text{ fails}) = 0,01 \\ P(K_2 \text{ fails}) = 0,03 \\ P(K_3 \text{ fails}) = 0,08 \end{array} \right\} \begin{array}{l} \text{can assume that} \\ \text{failures} \\ \text{mutually} \\ \text{independent} \end{array}$$

$$\begin{aligned} &\rightarrow 0,01 \cdot 0,03 \cdot 0,08 \\ &= 0,000024 \end{aligned}$$

$$\begin{aligned} P(\text{System does not fail}) \\ &= 1 - 0,000024 = 0,999976 \\ &= 99,99 \% // \end{aligned}$$

### Simulations 1.4.2

Real life situation hard to model exactly  $\rightarrow$  let computer run them

$n$  simulations, proportions where  $A$  happens :  $\hat{p}_n(A) \approx P(A)$

$$\text{Error-Estimation: } 1,96 \cdot \sqrt{\frac{\hat{p}_n(1-\hat{p}_n)}{n}}$$

$$\Rightarrow \hat{p}_n = \pm 1,96 \sqrt{\frac{\hat{p}_n(1-\hat{p}_n)}{n}}$$

(95% CONFIDENCE INTERVALS)

### Example 1.4.12

Person A: tosses coin

B: rolls die

repeated until H or 1,2,3,4

Compute  $P(A \text{ wins})$ :

Game completed if:

H or  $T \{1,2,3,4\}$  occurs

$$P(A) = \frac{\frac{1}{2}}{\frac{1}{2} + \frac{1}{2} \cdot \frac{4}{6}} = \frac{3}{5}$$

... let R-code run...

A wins in 0,6.