

Math 3235 Probability Theory

1/19/23

Bayes Theorem (Cont.) (Sec. 1.8)

Th. (Part.ition Th.)

If B_i form a partition of Ω and A is an event

$$P(A) = \sum_i P(A|B_i) P(B_i)$$

A : Test positive

B_1 : Healthy B_2 : sick

Ω is finite

$$B_i = \{i\} \quad P(B_i) = P(i)$$

$$A \cap \{i\} = \{i\} \quad \text{if } i \in A$$

$$= \emptyset \quad \text{if } i \notin A$$

$$P(A | B_i) = \begin{cases} 1 & \text{if } i \in A \\ 0 & \text{if } i \notin A \end{cases}$$

$$P(A) = \sum_i P(A|B_i) P(B_i) = \\ = \sum_{i \in A} P(i)$$

$$P(A) = P(A|B_1)P(B_1) + P(A|B_2)P(B_2)$$

We want To compute

$P(B_2 | A)$ one you know

$P(A | B_i)$ and $P(B_i)$

Th. (Bayes): If B_i form a partition and A is an event

$$P(B_i | A) = \frac{P(A | B_i) P(B_i)}{\sum_j P(A | B_j) P(B_j)}$$

Proof:

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

$$= \frac{P(A | B_i) P(B_i)}{P(A)}$$

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

↑ Partition Theorem



$$B_1 = \text{Healthy} = 99\%$$

$$B_2 = \text{sick} = 1\%$$

$A = \text{Test pos. True}$

$A^c = \text{The comp! set of } A =$

$$P(A | B_1) = 0.01 \quad -2 \setminus A.$$

$$P(A^c | B_1) = 0.99$$

$$P(A | B_2) = 1$$

$$P(A^c | B_2) = 0$$

$$P(B_2 | A) = \frac{P(A | B_2) P(B_2)}{P(A | B_1) P(B_1) + P(A | B_2) P(B_2)}$$

$$= \frac{1 \cdot 0.01}{0.01 \cdot 0.99 + 1 \cdot 0.01}$$

$$\approx 0.5$$

10.000

100 sick

199 positive

1.9 Probability are Continuous.

A_i are events

$A_1 \subset A_2 \subset A_3 \subset \dots \subset A_i \subset \dots$

$$P\left(\bigcup_{i=0}^{\infty} A_i\right) = \lim_{i \rightarrow \infty} P(A_i)$$

$$\left[\bigcup_{i=0}^{\infty} A_i = \lim_{i \rightarrow \infty} A_i \right]$$

Theorem.

If A_i is an increasing

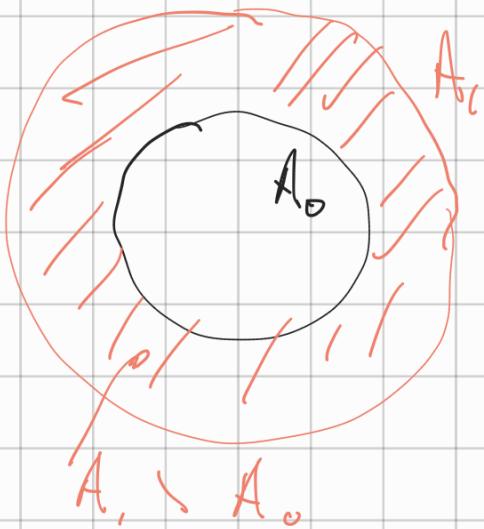
sequence of events Then

$$P\left(\bigcup_{i=0}^{\infty} A_i\right) = \lim_{i \rightarrow \infty} P(A_i)$$

$$\bigcup_{i=0}^n A_i = A_n$$

$$A_0 \cup A_1 = A_0 \cup (A_1 \setminus A_0)$$

$$B_i = A_{i+1} \setminus A_i$$



$$\bigcup_{i=0}^{\infty} A_i = A_0 \cup \bigcup_{i=0}^{\infty} B_i \quad B_i \cap B_j = \emptyset$$

$$P\left(\bigcup_{i=0}^{\infty} A_i\right) = P(A_0) + \sum_{i=0}^{\infty} P(B_i)$$

$$P(B_i) = P(A_{i+1}) - P(A_i)$$

$$P\left(\bigcup_{i=0}^{\infty} A_i\right) = P(A_0) + \lim_{n \rightarrow \infty} \sum_{i=0}^n P(A_{i+1}) - P(A_i)$$

$$(P(A_0) + (P(A_1) - P(A_0))) + (P(A_2) - P(A_1))$$

$$P\left(\bigcup_{i=0}^{\infty} A_i\right) = \lim_{n \rightarrow \infty} P(A_n)$$



Bayesian Statistics.

Bowl with n balls inside
 Some of the balls are red
 and some are blue.

Prior info: K number of
 red balls

$P(K)$: The probability that
 The number of red balls
 is K .

$P(K) = \frac{1}{n+1}$ for every K .

Someone extracts a ball and shows it to you: it's red.

Then he puts it back.

$g(k)$ probability that there k ball in the bag known the result of the extraction,

$$g(0) = 0$$

$$g(1) = P(K \text{ red balls} \mid \text{one was red})$$

$$= \frac{P(\text{red} \mid K \text{ red})}{\sum_{i=0}^n P(\text{red} \mid i \text{ red})} P(K \text{ red})$$

$$= \frac{\frac{k}{n}}{\sum_{i=0}^n \frac{i}{n}} \frac{\frac{1}{n+1}}{\sum_{i=0}^n \frac{i}{n+1}}$$

$$= \frac{K}{\sum_{i=0}^n i} = \frac{2K}{n(n+1)}$$