

## OMCIT 592 Module 08 Self-Paced 04 (instructor Val Tannen)

Reference to this self-paced segment in seg.08.06

This is a segment that contains material meant to be learned *at your own pace*. We are trying to assist you in this endeavor by organizing the material in a manner similar to the way it is outlined in the recorded segments, however with one additional suggestion.

When you see the following marker:



we suggest that you stop and make sure you thoroughly understood the material presented so far before you proceed further.

## Conditional probability rules

In the lecture segment “The chain rule” we stated the general property that justifies the calculation of probabilities in the tree of all possibilities by multiplying along a branch:

**Proposition (The Chain Rule).** For any **two or more** events  $A_1, \dots, A_n$  in the same probability space we have

$$\Pr[A_1 \cap A_2 \cap A_3 \cdots \cap A_n] = \Pr[A_1] \cdot \Pr[A_2 \mid A_1] \cdot \Pr[A_3 \mid A_1 \cap A_2] \cdots \Pr[A_n \mid A_1 \cap \cdots \cap A_{n-1}]$$

**Proof.** By induction on  $n$ .

(BC) The case  $n = 2$  follows directly from the definition of conditional probability:

$$\Pr[A_2 \mid A_1] = \frac{\Pr[A_1 \cap A_2]}{\Pr[A_1]} \quad \text{therefore} \quad \Pr[A_1 \cap A_2] = \Pr[A_1] \cdot \Pr[A_2 \mid A_1]$$

(IS) Let  $k \geq 2$  be an arbitrary natural number. Assume (IH) that

$$\Pr[A_1 \cap \cdots \cap A_k] = \Pr[A_1] \cdots \Pr[A_k \mid A_1 \cap \cdots \cap A_{k-1}]$$

We want to show that

$$\Pr[A_1 \cap \cdots \cap A_k \cap A_{k+1}] = \Pr[A_1] \cdots \Pr[A_k \mid A_1 \cap \cdots \cap A_{k-1}] \cdot \Pr[A_{k+1} \mid A_1 \cap \cdots \cap A_k]$$

Let  $B = A_1 \cap \cdots \cap A_k$ . As above, by the definition of conditional probability we have

$$\Pr[B \cap A_{k+1}] = \Pr[B] \cdot \Pr[A_{k+1} \mid B]$$

Using this, as well as the IH we obtain

$$\begin{aligned} \Pr[A_1 \cap \cdots \cap A_k \cap A_{k+1}] &= \Pr[B \cap A_{k+1}] \\ &= \Pr[B] \cdot \Pr[A_{k+1} \mid B] \\ &= \Pr[A_1] \cdots \Pr[A_k \mid A_1 \cap \cdots \cap A_{k-1}] \cdot \Pr[A_{k+1} \mid A_1 \cap \cdots \cap A_k] \end{aligned}$$



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## Conditional probability rules (continued)

**Problem (Rare Disease Test (again)).** A test for a disease that affects 0.1% of the population is 99% effective on people with the disease (i.e., it gives a **false negative** with probability 0.01). On people who do not suffer from the disease the test gives a **false positive** with probability 0.02. What is the probability that someone who tests positive in fact has the disease?

**Answer.** In the lecture segment “Conditional probability” we gave a detailed analysis of this problem using a tree of possibilities. Here, we give a different, faster, solution that illustrates two other useful properties of probability: Bayes’ Rule and the Total Probability Rule.

Again let  $V$  be a random person who takes the test. Let  $Y$  be the event that  $V$  has the disease and let  $+$  be the event that  $V$  tested positive. We wish to compute the conditional probability  $\Pr[Y | +]$ .

The problem gives us  $\Pr[Y] = 0.001$ .

We denote by  $N$  the complement of  $Y$ , which is the event that  $V$  does not have the disease.  $\Pr[N] = 0.999$ . The statement of the problem gives us the conditional probabilities for testing positive given that someone has, or that someone does not have the disease:

$$\Pr[+ | Y] = 0.99 \quad \Pr[+ | N] = 0.02$$

Note that we know  $\Pr[+ | Y]$  but we need  $\Pr[Y | +]$ . There is a property of probability that allows us to express one in terms of the other:

**Proposition (Bayes’ Rule).**

$$\Pr[A | B] = \frac{\Pr[A] \Pr[B | A]}{\Pr[B]}$$

There is not much to prove, as this follows directly from the definitions of conditional probabilities  $\Pr[A | B]$  and  $\Pr[B | A]$ .



Using Bayes’ Rule we get  $\Pr[Y | +] = \Pr[Y] \Pr[+ | Y] / \Pr[+]$  so we still need to compute  $\Pr[+]$  from what we know (note that we haven’t used  $\Pr[+ | N]$  yet).

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## Conditional probability rules (continued)

Again there is a useful property of probability that we can use:

**Proposition.** Let  $E$  and  $A$  be two events in the same probability space such that  $\Pr[A] \neq 0$ . Then

$$\Pr[E] = \Pr[E | A] \Pr[A] + \Pr[E | \bar{A}] \Pr[\bar{A}]$$

This is a particular case of a more general property called the Rule of Total Probability that we state without proof below. For our rare disease problem we need only this particular case so let's prove it.

**Proof.** Note that  $E \cap A$  and  $E \cap \bar{A}$  are disjoint. Note also (for example by drawing an Euler-Venn diagram) that  $(E \cap A) \cup (E \cap \bar{A}) = E$ . Therefore we can apply the addition rule (P2):

$$\Pr[E] = \Pr[E \cap A] + \Pr[E \cap \bar{A}]$$

By the definition of conditional probability

$$\Pr[E \cap A] = \Pr[E | A] \Pr[A] \quad \text{and} \quad \Pr[E \cap \bar{A}] = \Pr[E | \bar{A}] \Pr[\bar{A}]$$

and the proposition follows.



Applying this to our problem  $\Pr[+] = \Pr[+ | Y] \Pr[Y] + \Pr[+ | N] \Pr[N] = 0.99 \cdot 0.001 + 0.02 \cdot 0.999 = 0.00099 + 0.01998 = 0.02097$ . Then  $\Pr[Y | +] = \Pr[Y] \Pr[+ | Y] / \Pr[+] = 0.00099 / 0.02097 \simeq 0.0472$ , the same answer we obtained in our first solution that used a tree of all possibilities (in the lecture segment “Conditional probability”).



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## Conditional probability rules (continued)

As promised we state without proof (the proof is by induction, not hard, but somewhat tedious):

**Proposition (The Rule of Total Probability).** Let  $A_1, \dots, A_n$  be **two or more** events, each of **non-zero probability** in the same probability space such that:

- $A_1, \dots, A_n$  are **pairwise disjoint**, and
- $A_1 \cup \dots \cup A_n = \Omega$

(we say that  $A_1, \dots, A_n$  form a **partition** of  $\Omega$ ). Then, for any event  $E$

$$\Pr[E] = \sum_{i=1}^n \Pr[E \mid A_i] \Pr[A_i]$$

