

## MDI Probability Theory.

### Class 4: Total Probability. Bayes rule.

#### Total probability (Partition theorem)

Let  $(\Omega, \mathcal{F}, P)$  be a probability space for a particular random experiment. We introduce term *partition* — collection  $\{B_i\}$  of pairwise disjoint events  $B_i$ , such that  $\bigcup_i B_i = \Omega$ .

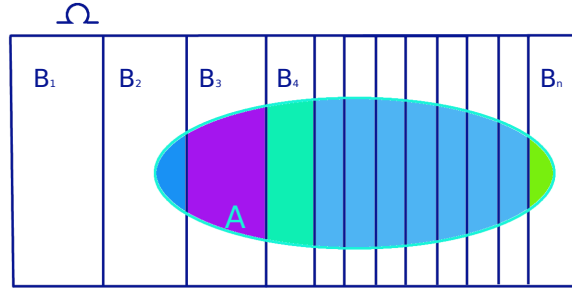
The following result, based on partition concept, is of great importance:

**Theorem:** If  $\{B_i\}$  is a partition of  $\Omega$ , with  $P(B_i) > 0$ ,  $\forall i$ , then for any event  $A$  from  $\mathcal{F}$ :

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

Indeed, let us reconstruct event  $A$  from its intersections with all events  $B_i$ :

$$A = \bigcup_i (A \cap B_i)$$



Initial events  $\{B_i\}$  are pairwise disjoint, so their intersections  $\{(A \cap B_i)\}$  with  $A$  are also pairwise disjoint, because each intersection is like restriction for  $B_i$  — there are no new elements. Since that, we can apply important property (additivity) of the probability function:

$$P(A) = P\left(\bigcup_i (A \cap B_i)\right) = \sum_i P(A \cap B_i).$$

After that let's apply concept of conditional probability:

$$P(A \cap B_i) = P(A|B_i)P(B_i),$$

and collect everything together:

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

The intuition is that event  $A$  may happen as a consequence of some others events  $\{B_i\}$  which together completely deplete the sample space ( $\bigcup_i B_i = \Omega$ ). In this case we can decompose probability of  $A$  to the sum of probabilities of initial  $B_i$ 's multiplied by conditional probability of  $A$  to happen if  $B_i$  has happened.

One of the common situations is when partition is represented by some event  $B$  and its complement:  $\{B, B^c\}$ . This means that event  $A$  may happen only after either event  $B$  or  $B^c$ .

Example: Tomorrow there will be either rain or snow but not both; the probability of rain is  $\frac{2}{5}$  and the probability of snow is  $\frac{3}{5}$ . If it rains, the probability that I will be late for my lecture is  $\frac{1}{5}$ , while the corresponding probability in the event of snow is  $\frac{3}{5}$ . What is the probability that I will be late?

Let  $A$  be the event that I am late and  $B$  be the event that it rains. The collection  $\{B, B^c\}$  is a partition of the sample space. Then, by total probability principle we have:

$$P(A) = P(A|B)P(B) + P(A|B^c)P(B^c) = \frac{1}{5} \cdot \frac{2}{5} + \frac{3}{5} \cdot \frac{3}{5} = \frac{11}{25}.$$

### Bayes rule

Sometimes we can treat event  $A$  as some observation, or evidence, and events  $B_i$  as some states of nature which precede  $A$ . Clearly as  $B_i$ 's are disjoint, there is only one  $B_i$  happening together with  $A$ . If probabilities  $P(A|B_i)$  are known, we may want to have an answer to the question "How likely that specifically this  $B_i$  preceded  $A$ ?", *i.e.* to find a probability  $P(B_i|A)$ .

**Theorem (Bayes rule):** Let  $\{B_i\}$  is a partition of  $\Omega$ , with  $P(B_i) > 0$ ,  $\forall i$ . Then for any event  $A$ :

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

Let us recall definition of conditional probability:

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

We can notice that probability of intersection ( $A \cap B$ ) may be written in two ways:

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

which gives us a formula, how two 'inverted' conditional probabilities are connected:

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

Let us change  $B$  by some element of partition  $B_j$ , and use the total probability formula for  $P(A)$ , and we obtain needed result.

Example (**False positives**): A rare but potentially fatal disease has an incidence of 1 in  $10^5$  in the population at large. There is a diagnostic test, but it is imperfect. If you have the disease, the test is positive with probability  $\frac{9}{10}$ ; if you do not, the test is positive with probability  $\frac{1}{20}$ . Your test result is positive. What is the probability that you have the disease?