

# W02 - Examples

## Bayes' Theorem

### Example - Bayes' Theorem - COVID tests

Assume that 0.5% of people have COVID. Suppose a COVID test gives a (true) positive on 96% of patients who have COVID, but gives a (false) positive on 2% of patients who do not have COVID. Bob tests positive. What is the probability that Bob has COVID?

#### Solution

##### 1. Label events.

- Event  $A_P$ : Bob is actually positive for COVID
- Event  $A_N$ : Bob is actually negative; note  $A_N = A_P^c$
- Event  $T_P$ : Bob tests positive
- Event  $T_N$ : Bob tests negative; note  $T_N = T_P^c$

##### 2. Identify knowns.

- Know:  $P[T_P | A_P] = 96\%$
- Know:  $P[T_P | A_N] = 2\%$
- Know:  $P[A_P] = 0.5\%$  and therefore  $P[A_N] = 99.5\%$
- We seek:  $P[A_P | T_P]$

##### 3. Translate Bayes' Theorem.

- Using  $A = T_P$  and  $B = A_P$  in the formula:

$$P[A_P | T_P] = P[T_P | A_P] \cdot \frac{P[A_P]}{P[T_P]}$$

- We know all values on the right except  $P[T_P]$

##### 4. Use Division into Cases.

- Observe:

$$T_P = T_P \cap A_P \cup T_P \cap A_N$$

- Division into Cases yields:

$$P[T_P] = P[A_P] \cdot P[T_P | A_P] + P[A_N] \cdot P[T_P | A_N]$$

- Important to notice this technique!

- It is a common element of Bayes' Theorem application problems.
- It is frequently needed *for the denominator*.

- Plug in data and compute:

$$\gg \gg P[T_P] = \frac{5}{1000} \cdot \frac{96}{100} + \frac{995}{1000} \cdot \frac{2}{100} \gg \gg \approx 0.0247$$

##### 5. Compute answer.

- Plug in and compute:

$$P[A_P | T_P] = P[T_P | A_P] \cdot \frac{P[A_P]}{P[T_P]}$$

$$\gg \gg \quad 0.96 \cdot \frac{0.005}{0.0247} \quad \gg \gg \quad \approx 19\%$$

### 🔗 Intuition - COVID testing

Some people find the low number surprising. In order to repair your intuition, think about it like this: roughly 2.5% of tests are positive, with roughly 2% coming from *false* positives, and roughly 0.5% from *true* positives. The true ones make up only 1/5 of the positive results!

(This rough approximation is by assuming 96% = 100%.)

If *two* tests both come back positive, the odds of COVID are now 98%.

If *only people with symptoms* are tested, so that, say, 20% of those tested have COVID, that is,  $P[A_P | T_P] = 20\%$ , then one positive test implies a COVID probability of 92%.

### 🔗 Exercise - Bayes' Theorem and Multiplication: Inferring bin from marble

There are marbles in bins in a room:

- Bin 1 holds 7 red and 5 green marbles.
- Bin 2 holds 4 red and 3 green marbles.

Your friend goes in the room, shuts the door, and selects a random bin, then draws a random marble. (Equal odds for each bin, then equal odds for each marble in that bin.) He comes out and shows you a red marble.

What is the probability that this red marble was taken from Bin 1?

[Solution](#)

## Independence

### 🔗 Exercise - Independence and complements

Prove that these are logically equivalent statements:

- $A$  and  $B$  are independent
- $A$  and  $B^c$  are independent
- $A^c$  and  $B^c$  are independent

Make sure you demonstrate both directions of each equivalency.

Solution**Example - Checking independence by hand**

A bin contains 4 red and 7 green marbles. Two marbles are drawn.

Let  $R_1$  be the event that the first marble is red, and let  $G_2$  be the event that the second marble is green.

- (a) Show that  $R_1$  and  $G_2$  are independent if the marbles are drawn *with replacement*.
- (b) Show that  $R_1$  and  $G_2$  are not independent if the marbles are drawn *without replacement*.

**Solution**

(a) With replacement.

1. **Identify knowns.**

- Know:  $P[R_1] = \frac{4}{11}$
- Know:  $P[G_2] = \frac{7}{11}$

2. **Compute both sides of independence relation.**

- Relation is  $P[R_1 G_2] = P[R_1] \cdot P[G_2]$
- Right side is  $\frac{4}{11} \cdot \frac{7}{11}$
- For  $P[R_1 G_2]$ , have  $4 \cdot 7$  ways to get  $R_1 G_2$ , and  $11^2$  total outcomes.
- So left side is  $\frac{4 \cdot 7}{11^2}$ , which equals the right side.

(b) Without replacement.

1. **Identify knowns.**

- Know:  $P[R_1] = \frac{4}{11}$  and therefore  $P[R_1^c] = \frac{7}{11}$
- We seek:  $P[G_2]$  and  $P[R_1 G_2]$

2. **Find  $P[G_2]$  using Division into Cases.**

- Division into cases:

$$G_2 = G_2 \cap R_1 \cup G_2 \cap R_1^c$$

- Therefore:

$$P[G_2] = P[R_1] \cdot P[G_2 | R_1] + P[R_1^c] \cdot P[G_2 | R_1^c]$$

- Find these by counting and compute:

$$\gg \gg P[G_2] = \frac{4}{11} \cdot \frac{7}{10} + \frac{7}{11} \cdot \frac{6}{10} \gg \gg \frac{70}{110}$$

3. **Find  $P[R_1 G_2]$  using Multiplication rule.**

- Multiplication rule (implicitly used above already):

$$P[R_1 G_2] = P[R_1] \cdot P[G_2 | R_1] = \frac{4}{11} \cdot \frac{7}{10} = \frac{28}{110}$$

4. **Compare both sides.**

- Left side:  $P[R_1 G_2] = \frac{28}{110}$

- Whereas, right side:

$$P[R_1] \cdot P[G_2] = \frac{4}{11} \cdot \frac{70}{110} = \frac{28}{121}$$

- But  $\frac{28}{110} \neq \frac{28}{121}$  so  $P[R_1 G_2] \neq P[R_1] \cdot P[G_2]$  and they are *not independent*.