

1. Introduction to Bayes' Theorem

Bayes' theorem is a useful tool to calculate conditional probability of an event; that is, to calculate probability of an event based on prior knowledge or additional information related to an event.

For example, when rolling a dice, the probability of getting a '6' is 1/6. If I tell you that the result of the roll is an even number, what is the probability of getting a '6'? Obviously it is much higher than 1/6. This simple example illustrates the importance of **additional information or prior knowledge**.

2. Simple Derivation of Bayes' Theorem

We denote $P(A|B)$ as the probability of A will happen given that B **already happened**; that is, $P(A|B) = P(A \text{ Will Happen} | B \text{ Already Happened})$

$P(A|B)$ can be calculated using the following:

$$P(A|B) = \frac{P(A \cap B)}{P(B)} \quad (\text{Equation 1})$$

Where $P(A \cap B)$ stands for A and B happen at the same time.

Using the same reasoning, we can obtain:

$$P(B|A) = \frac{P(B \cap A)}{P(A)} \quad (\text{Equation 2})$$

Since $P(B \cap A) = P(A \cap B)$, by substituting 2nd Equation to the 1st equation, the following expression can be obtained:

$$P(A|B) = \frac{P(B|A)P(A)}{P(B)} \quad (\text{Equation 3})$$

Equation 3 is the formula for Bayes' theorem!

Sometimes $P(B)$ is not given for a problem. $P(B)$ is the unconditional probability that B will happen, or put it in another way, $P(B)$ is the probability that B will happen no matter A happens or not.

Thus, $P(B)$ can be expanded to:

$$P(B) = P(B \cap A) + P(B \cap A^c) \quad (\text{Equation 4})$$

where A^c indicate A does not happen. After plugging equation 2 to equation 4, we can obtain:

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

$$P(B \cap A^c) = P(B|A^c) * P(A^c)$$

Then Equation 3 becomes the following, which is another way to express Bayes Theorem:

$$P(A|B) = \frac{P(B|A)P(A)}{P(B|A) * P(A) + P(B|A^c) * P(A^c)} \quad (\text{Equation 5})$$

3-1 Example 1

Calculate the probability of getting a '6' after rolling a dice, given that the result (R) is an even number.

Let's show the formular for Baye's theorem again:

$$P(A|B) = \frac{P(B|A)P(A)}{P(B)} \quad (\text{Equation 3})$$

In here we want to calculate:

$P(A|B) = P(\text{rolling a 6} \mid \text{result is even})$

Information We Know:

$P(A) = P(6) =$ the *unconditional* probability of getting a '6' = $1/6$

$P(B) = P(\text{even}) =$ the *unconditional* probability of getting an even number = $1/2$

$P(B|A) =$ probability of getting an even number aftering rolling a '6' = 1

Thus:

$$P(6|\text{even}) = \frac{P(\text{even}|6)P(6)}{P(\text{even})} = \frac{1 * (1/6)}{1/2} = 1/3$$

3-2 Example 2

Suppose I have two children and assume that the probability of giving birth to a boy and a girl are identical. Given that one of my children is a girl, waht is the probability of both children are girls.

In here we want to calculate:

$P(A|B) = P(\text{both are girls} \mid \text{one of the children is girl})$

Information We Know:

$P(A) = P(\text{both girls}) = 1/4$ $P(B) = P(\text{one of the children is girl}) = 3/4$ $P(B|A) = P(\text{one of the children is girl} \mid \text{both are girls}) = 1$

Answer:

$P(A|B) = P(\text{both girls} \mid \text{one of children is girl}) = P(\text{one of children is girl} \mid \text{both are girls}) * P(\text{both girls}) / P(\text{one of children is girl}) = 1 * (1/4) / (3/4) = 1/3$

3-3 Example 3

Monty Hall Problem:

Suppose you're on a game show, and you're given the choice of three doors: Behind one door is a car; behind the other two, goats.

You pick a door, say Door A, and the host, who knows what's behind the doors, opens another door, say Door B, which has a goat. He then says to you, "Do you want to pick Door C or stick with Door A?" **Is it to your advantage to switch your choice?**

In here we want to calculate:

$P(A|B) = P(\text{Car @ C} | \text{Open Door B})$. The probability of the car being at Door C (the door we did not choose) given that the host tell us the car is not at Door B.

Information We Know:

$P(A) = P(\text{Car @ C}) = 1/3$, since only 1 out of 3 doors contains the car

$P(B|A) = P(\text{Open Door B} | \text{car @ C}) = 1$ since if the car is at Door C, then the car cannot be at Door B

Calculating $P(B)$, which is the probability of the host opening Door B is a little tricky because it is not obvious in the problem. Suppose you selected Door A, then the host will open Door B when the Car is either at Door A or Door C, giving the following relationship:

$P(B) = P(\text{Open Door B}) = P(\text{Open Door B} \cap \text{Car @ A}) + P(\text{Open Door B} \cap \text{Car @ B}) + P(\text{Open Door B} \cap \text{Car @ C})$. Using the conditional probability formula we can obtain:

$P(\text{Open Door B} \cap \text{Car @ A}) = P(\text{Open Door B} | \text{Car @ A}) * P(\text{Car @ A}) = (1/2)*(1/3) = 1/6$. In here, $P(\text{Open Door B} | \text{Car @ A})$ is $1/2$ because the host can open either Door B and Door C because they are both goat.

$P(\text{Open Door B} \cap \text{Car @ B}) = 0$, since the host will never reveal the location of the car

$P(\text{Open Door B} \cap \text{Car @ C}) = P(\text{Open Door B} | \text{Car @ C}) * P(\text{Car @ C}) = 1*(1/3) = 1/3$. In here, $P(\text{Open Door B} | \text{Car @ C}) = 1$ because Door A is occupied, Door C has car, and the host must open door B

As a result, $P(B) = P(\text{Open Door B}) = 1/6 + 0 + 1/3 = 1/2$

Answer:

$P(A|B) = P(\text{Car @ C} | \text{Open Door B}) = P(\text{Open Door B} | \text{car @ C}) * P(\text{Car @ C}) / P(\text{Open Door B}) = 1 * (1/3) / (1/2) = 2/3$. This is the probability of getting a car after **Switching** from door A to door C.

What if we stick with Door A even after the host reveal Door B is a goat?

$P(\text{Car @ A} | \text{Open Door B}) = P(\text{Open Door B} | \text{car @ A}) * P(\text{car @ A}) / P(\text{Open Door B}) = (1/2) * (1/3) / (1/2) = 1/3$.

This is the probability of getting a car after **Sticking** with Door A.

The probability of winning a car after switching is twice as much as not switching; as a result, switching is always a good strategy.

3-4 Example 4

Suppose there is a device that detect whether someone has a disease with a accuracy of 99%. That is, if there are 100 sick people, then the device will produce Test Positive result for 99 out of 100 people. If there are 100 healthy people, then the device will produce Test Negative for 99 out of 100 people.

After checking the population statistics, we found that only 1% of the people on earth have this disease.

Suppose a patient got a test positive for the disease. What is the probability that the person is really sick?

In here we want to calculate:

$P(A|B) = P(\text{Sick} | \text{Test Positive})$. The probability of really being sick given that the device is telling the patient he/she is sick.

Information We Know:

$P(A) = P(\text{Sick}) = 1\%$ since only 1% of the population have this disease $P(B|A) = P(\text{Test Positive} | \text{Sick}) = 99\%$

$P(B) = P(\text{Test Positive})$. This is the unconditional probability of test positive, which is the probability of getting a Test Positive Answer despite of the patient is really sick or not. It can be calculated by the following:

$P(B) = P(\text{Test Positive}) = P(\text{Test Positive} \cap \text{Sick}) + P(\text{Test Positive} \cap \text{healthy})$.

$P(\text{Test Positive} \cap \text{Sick}) = P(\text{Test Positive} | \text{Sick}) * P(\text{Sick}) = 99\% * 1\% = 0.99\%$

$P(\text{Test Positive} \cap \text{Healthy}) = P(\text{Test Positive} | \text{Healthy}) * P(\text{Healthy}) = 1\% * 99\% = 0.99\%$

Thus, $P(B) = 0.99\% + 0.99\% = 1.98\%$

Answer:

$P(A|B) = P(\text{Sick} | \text{Test Positive}) = P(\text{Test Positive} | \text{Sick}) * P(\text{Sick}) / P(\text{Test Positive}) = 99\% * 1\% / 1.98\% = 50\%$

The probability of being sick given that the device produce Test Positive result is 50%