

# BAYES RULE

## For the Evidence of Testimony

The conditional probability (Prob[E | A]) in this case is that an event (E; a miracle) took place, given that the witness said it took place (A). Bayes formula in this case requires the prior probabilities, Prob[E] and Prob[~E], leading to the formula:

$$Prob[E | A] = \frac{Prob[E] Pr[A | E]}{(Prob[E] Prob[A | E]) + (Prob[~E] Prob[A | ~E])}$$

Observe, Prob[A | E] -- that the witness said it took place and the event did take place -- is the probability that the witness is truthful; and Prob[A | ~E] = (1 - Prob[A | E]) is the probability the witness is not truthful.

**MIRACLES** Did the Christian miracles occur?

### EXAMPLE 1

Prob[E] = Prob. of Miracle = .01      Prob[~E] = Prob. of no miracle = .99  
Prob[A | E] = Prob. of truthful witness = .75      Prob[A | ~E] = .25

$$Prob[E | A] = \{(.01)(.75)\} / \{(.01)(.75) + (.99)(.25)\} = .0294$$

### EXAMPLE 2

At one point, Price argues that Hume is incorrect to allow prior probabilities to impact the judgment of whether an event happened. In this case:

Prob[E] = Prob. of Miracle = Prob[~E] = Prob. of no miracle = .5  
Prob[A | E] = Prob. of truthful witness = .99      Prob[A | ~E] = .01

$$Prob[E | A] = .99$$

### EXAMPLE 3    Prob[E] = 1 - Prob[A | E]

Prob[E] = Prob. of Miracle = .01      Prob[~E] = Prob. of no miracle = .99  
Prob[A | E] = Prob. of truthful witness = .99      Prob[A | ~E] = .01

$$Prob[E | A] = .5$$

## GENERALIZING BAYES

Bayes Rule is more general than the miracle/no miracle two alternative case. When there are more than two mutually exclusive alternatives of which one must occur and none has zero probability, the denominator can be further expanded as:

$$Prob[A] = \sum_{i=1}^k Prob[B_i] prob[A | B_i]$$

This produces the following generalization:

$$Prob[B_j | A] = \frac{Prob[B_j] Prob[A | B_j]}{\sum_{i=1}^k Prob[B_i] Prob[A | B_i]} = \frac{Prob[B_j] Prob[A | B_j]}{Prob[A]}$$

**PARTIES:** Did Tom have a 'good time'?

Having discovered that Tom had a good time at the party (event F happened), what is the probability that his ex-wife did not attend the party (Prob[B1 | F])?

Prob[Tom's ex-wife goes to the Party] = Prob[Tom's ex-wife does not go to the Party] = .5  
Prob[B2] = Prob[B1]

Note: This is .5 because there is no information about wife's attendance at the party, i.e., B2 = ~B1 and the probabilities will sum to one.

Prob.[Tom has a good time if wife goes to party] = .2 = Prob[F | B2]  
Prob.[Tom has a good time if wife does not go to party] = .75 = Prob[F | B1]

Note: These two probabilities do not sum to one, i.e., it is Prob[F | B1] + Prob[~F | B1] = 1

Solution:

$$\begin{aligned} Prob[B1 | F] &= \{Prob[B1] Prob[F | B1]\} / \{(Prob[B1] Prob[F | B1]) + \{Prob[B2] Prob[F | B2]\}\} \\ &= (.5)(.75) / \{(.5)(.2) + (.5)(.75)\} = 0.79 \end{aligned}$$

What if it is certain that Tom will not have a good time if ex-wife attends, i.e., Prob[F | B2] = 0? Observe that in this (trivial) case, it is also certain the ex-wife did not attend → plug into Bayes theorem to verify.