

## Bayesian inference

The main idea of Bayesian analysis is easy to state. Suppose  $p$  is the proportion of voters in Florida who intend to vote for Bush. Note that this isn't quite the same  $p$  that appeared earlier; that  $p$  was a *probability*, not a proportion. In this section, if I say  $p = .5$  I mean that exactly half of the votes go to Bush, and the election is a tie.

Using the binomial theorem, we find that the probability of obtaining 366 votes for Bush and 303 for Kerry in our sample of 669 voters is

$$\binom{669}{303} p^{366} (1-p)^{303}.$$

We can thus compute that, if  $p = .55$ , the chance that a random sample of 669 voters would contain 336 Bush supporters is about 3 percent. If  $p = .5$ , we should expect it to be less likely that the sample will come out so strongly in favor of Bush; and indeed, we find that when  $p = .5$  the chance of obtaining a 366 – 303 margin in our sample is just 0.0016, about 20 times less likely. Bayes's theorem allows us to turn this analysis on its head, saying: given that the poll came out 366 – 303, it is 20 times more likely that  $p = .55$  than that  $p = .5$ . Applying this argument to all possible values of  $p$  eventually allows us to specify precisely the probability that  $p$  assumes any particular value. Now if you don't want to see equations, don't read any further!

But if you do: suppose  $A$  and  $B$  are two events, and let  $P(A)$  be the probability that  $A$  occurs,  $P(B)$  the probability that  $B$  occurs,  $P(A, B)$  the probability that both  $A$  and  $B$  occur, and  $P(A|B)$  the probability that  $A$  occurs given that  $B$  occurs. Then Bayes's theorem says

$$P(A|B) = P(A, B)/P(B).$$

In this case, we can take  $B$  to be the event "The poll results are 366-303 in favor of Bush" and  $A$  to be the event "The voters in Florida are evenly split, 3 million for Bush and 3 million for Kerry." What we are trying to compute is  $P(A|B)$ ; the probability that the vote is a tie, given the 366-303 poll result. What we *know* about is  $P(B|A)$ , the probability that we'll obtain a 366-303 poll result given equal numbers of Bush and Kerry voters in Florida; this probability, as mentioned above, is about 0.0016.

For each number  $n$  between 0 and 6000000, write  $A_n$  for the event "The proportion of voters in Florida supporting Bush is  $n$ ." Then we've already seen that

$$P(B|A_n) = \binom{669}{303} p^{366} (1-p)^{303}.$$

where  $p = n/6000000$  is the proportion of voters supporting Bush. And, since we've stipulated that our prior belief holds all values of  $n$  to be equally likely, we can say  $P(A_n) = 1/6000000$  for all  $n$ ; so by Bayes's theorem

$$P(B, A_n) = P(B|A_n)P(A_n) = (1/6000000)P(B|A_n).$$

Now I claim

$$P(B) = \sum_{n=0}^{6000000} P(B, A_n);$$

this amounts to saying “the chance that the poll comes out 366-303 is obtained by summing, over all possible  $n$ , the chance that the poll comes out 366 – 303 and that  $n$  voters in Florida support Bush.” This sum is just

$$(1/6000000) \binom{699}{303} \sum_{n=0}^{6000000} (n/6000000)^{366} (1 - n/6000000)^{303}$$

which comes out to about 0.0149. Now we can return to our original problem: using Bayes’s theorem again, we have

$$P(B|A) = P(B, A)/P(A) = 6000000P(B, A).$$

Whence

$$P(A|B) = P(B, A)/P(B) = P(B|A)/(6000000P(B)).$$

Since  $P(B|A) = 0.0016$  and  $P(B) = 0.0149$ , we conclude that

$$P(A|B) = 1.77 \times 10^{-7}$$

which is to say that, given the 366-303 poll result, the chance of a perfect tie is about 1 in 5000000, as claimed.