

# Conditional Probability and Likelihood

A random variable is defined by the result of a rule (e.g., a function) that associates a real number with each outcome in a sample space,  $S$ . Consider two discrete random variables,  $A$  and  $B$ , that are associated with events in the sample space. The probabilities of  $A$  and  $B$ , denoted by  $P(A)$  and  $P(B)$ , respectively, take values between zero and one and, by definition,  $P(S) = 1$ . Conditional probability is denoted by  $P(A | B)$  and read as “the probability of  $A$  given  $B$ .” Given that  $B$  has occurred, the sample space shrinks from  $S$  to  $B$ .

We expect that  $P(A | B)$  is proportional to  $P(A \cap B)$ , which is the intersection probability that both  $A$  and  $B$  occur. Since  $P(B | B)$  must be unity and  $P(B \cap B) = P(B)$ , the constant of proportionality must be  $1/P(B)$ :

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

Similarly,

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

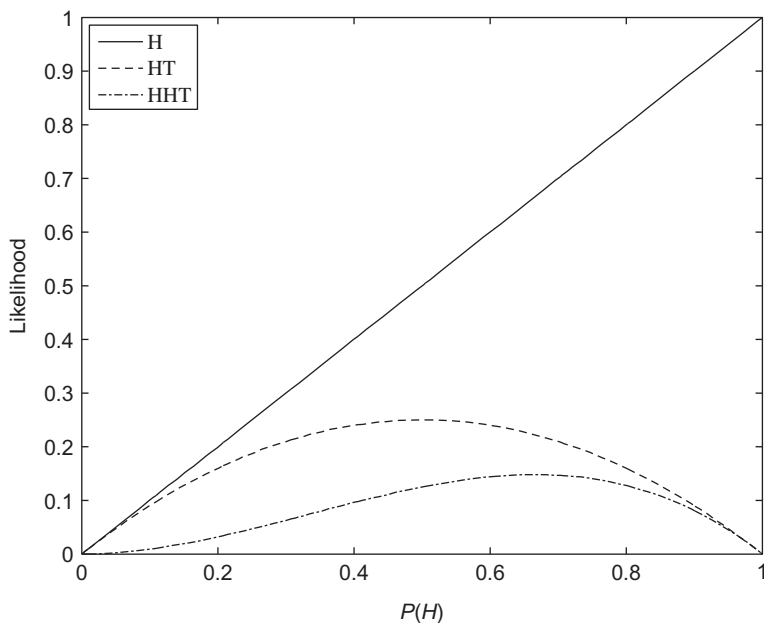
Using (G.2) to replace  $P(A \cap B)$  in (G.1) gives **Bayes’s Theorem**,

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

## Likelihood

With conditional probability, we are given specific information about an event that affects the probability of another event. For example, knowing that  $B = b$  allows us to “update” the probability of  $A$  by  $P(A | B = b)$  through (G.3). Conversely, if we view the conditional probability as a function of the *second* argument,  $B$ , we get a likelihood function

$$L(b | A) = \alpha P(A | B = b) \quad (\text{G.4})$$



**FIGURE G.1** Likelihood as a function of the probability of “H” on a two-sided coin given observations of heads “H” and tails “T.”

where  $\alpha$  is a constant parameter. For example, consider a sequence of observations of a two-sided coin with outcomes “H” and “T.” If we flip the coin once and observe “H,” we can ask, “What is the likelihood that the true probability  $P(H)$  is 0.5?” Figure G.1 shows that  $L = 0.5$ , and in fact the most likely value is  $P(H) = 1$ ; note that  $L(0) = 0$  since “H” has been observed and therefore must have a nonzero probability. Suppose flipping the coin again also gives “H.” Maximum likelihood is still located at  $P(H) = 1$ , and the likelihood of smaller  $P(H)$  is diminished. Remember that we can turn this around and ask, “If we *know* that  $P(H) = 0.5$  (i.e., the coin is fair), what is the probability of observing ‘HH’?” This is conditional probability, which in this case is 0.25 (the same as  $L(0.5 | HH)$ ). If, on the other hand, one of the observations is also “T,” then both  $L(0)$  and  $L(1)$  must be zero (Figure G.1). Note that, unlike conditional probability likelihood is not a probability since it may take values larger than unity and it does not integrate to one.