1 Probability and Likelihood [3]

1.1 Definition

- Probabilities defined over discrete sets of events: $Prob(dice = 1 | dice \in \{1, 2, 3, 4, 5, 6\}) = 1/6$.
- Probabilities with respect to continuous variables: $\text{Prob}(X = 5.0 | X \in [1.0, 6.0]) = 1/\infty = 0.$
 - Probability density function (PDF): y = f(x). $Prob(x \in [2.0, 4.0]|f(x) = 0.2$, for $1.0 \le x \le 6.0$) = $\int_{2.0}^{4.0} f(x) dx = 0.2 \times 2 = 0.4$, where f(x) is shown in Fig. 1(a).

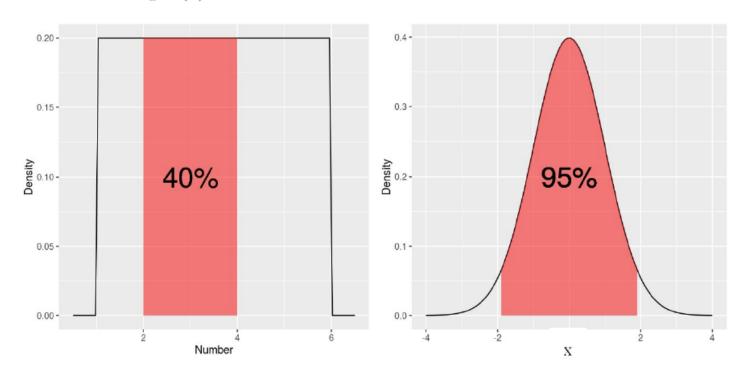


Figure 1: Probability density functions (PDFs) of continuous variables: (a) uniform distribution in [1.0, 6.0], y = 0.2, for $1.0 \le x \le 6.0$ and (b) standard normal distribution, $y = \frac{1}{\sqrt{2\pi}}e^{-x^2/2}$

1.2 Maximum Likelihood Estimation

Given an experiment (event or system), estimate the parameter (or a set of parameters) that maximizes the likelihood of the experiment.

Example 1. Estimation of the probability of an uneven coin using multiple trials

Let p be an unknown probability of head when tossing an uneven coin, then the probability of tail is equal to 1-p. To estimate p, that is considered as a parameter of the coin tossing system, we had the head 400 times and tails 600 times out of 1000 trials. The estimation problem is illustrated in Fig. 2.

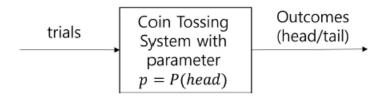


Figure 2: Estimation of uneven coin probability

The estimation process includes the following steps:

- 1. Get outcomes by trials. (400 times head, 600 times tail)
- 2. Define the likelihood as a function of p, such as $L = {}_{1000}C_{400}p^{400}(1-p)^{600}$ as shown in Fig. 3.

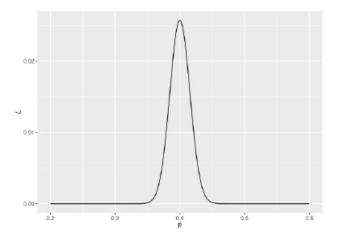


Figure 3: Likelihood function, $L = {}_{1000}C_{400}p^{400}(1-p)^{600}$

3. To maximize L, $\frac{dL}{dp} = 0$ yields the solution. (p = 0.4)

The estimated parameter maximizes the likelihood of given outcomes. In a continuous case, the likelihood is equivalent to the probability density function of the outcome, which states that p = 0.4 maximizes the probability of the outcome (400 head, 600 tail).