

Given is a realization (x_1, \dots, x_n) of samples of $n = 100$ coin flips. The values $x_i = 1$ represent heads and $x_i = 0$ tails. We are interested in estimate the unknown parameter θ that is associated with the probability of heads. In order to approximate θ consider two different statistical models:

- it is assumed for the statistical model that the underlying X_1, \dots, X_n are independent and identical distributed random variables following the Bernoulli distribution
- it is assumed for the statistical model that the underlying X_1, \dots, X_n are independent and identical distributed random variables following the Binomial distribution.

Estimate θ using the Maximum Likelihood Method for the two statistical models. Compare the resulting values and comment on the difference.

Solution:

If a random sample x_1, \dots, x_n of n observations is drawn from a Bernoulli distribution with parameter θ , this leads to the following likelihood and log-likelihood functions [1]:

$$L(\theta) = (1 - \theta)^{(1-x_1)} * \theta^{x_1} \dots \dots \dots (1 - \theta)^{(1-x_n)} * \theta^{x_n} = \prod_{i=1}^n (1 - \theta)^{(1-x_i)} * \theta^{x_i} \dots \dots \dots (1)$$

Now, in the “sampleset.txt” the number of heads (i.e., 1) is 43. So, we can write the equation (1) as follows:

$$L(\theta) = (1 - \theta)^{(100-43)} * \theta^{43} \dots \dots \dots (2)$$

Take the log of (2), we got:

$$\ln(L(\theta)) = \ln((1 - \theta)^{57} * \theta^{43}) = 43 * \ln \theta + 57 * \ln(1 - \theta) \dots \dots \dots (3)$$

Take the derivative of the likelihood function equation (2) and setting it to 0, we found,

$$\begin{aligned} \frac{d}{d\theta} L(\theta) &= \frac{d}{d\theta} (43 * \ln \theta + 57 * \ln(1 - \theta)) \\ &= \frac{43}{\theta} - \frac{57}{1 - \theta} \end{aligned}$$

As $\frac{d}{d\theta} L(\theta) = 0$ so:

$$\begin{aligned} \frac{43}{\theta} - \frac{57}{1 - \theta} &= 0 \\ \Rightarrow \frac{43 * (1 - \theta) - 57 * \theta}{\theta * (1 - \theta)} &= 0 \\ \Rightarrow 43 - 43 * \theta - 57 * \theta &= 0 \\ \Rightarrow 100 * \theta &= 43 \\ \Rightarrow \theta &= \frac{43}{100} \\ \Rightarrow \theta &= 0.43 \end{aligned}$$

Assume that X is a binomial distribution observation, $X \sim \text{Bin}(n, \theta)$, where n is known and θ is to be estimated. The probability function is as follows [2]:

$$L(x; \theta) = \frac{n!}{x! * (n-x)!} * \theta^x * (1 - \theta)^{n-x}$$

Now [3],

$$\begin{aligned} L(x_1, \dots, x_{100}; \theta) &= P_{X_1 \dots X_{100}}(x_1, \dots, x_{100}; \theta) \\ &= P_{X_1}(x_1; \theta) \dots P_{X_{100}}(x_{100}; \theta) \\ &= \binom{100}{1}^{43} * \binom{100}{0}^{57} * \theta^{43} * (1 - \theta)^{57} \text{ [given that the number of head is 43} \\ &\text{and tail is 57]} \\ &= 1 \times 10^{86} * \theta^{43} * (1 - \theta)^{57} \end{aligned}$$

We may take the derivative and set it to zero to discover the value of θ that maximizes the likelihood function. We have got:

$$\begin{aligned} \frac{d}{d\theta} L(x_1, \dots, x_{100}; \theta) &= 1 \times 10^{86} * (43 * \theta^{42} * (1 - \theta)^{57} - 57 * \theta^{43} * (1 - \theta)^{56}) = 0 \\ \Rightarrow 43 * \theta^{42} * (1 - \theta)^{57} - 57 * \theta^{43} * (1 - \theta)^{56} &= 0 \\ \Rightarrow \frac{43}{57} * \frac{(1 - \theta)^{57}}{(1 - \theta)^{56}} &= \frac{\theta^{43}}{\theta^{42}} \\ \Rightarrow \frac{43}{57} * (1 - \theta) &= \theta \\ \Rightarrow \frac{43}{57} - \frac{43 * \theta}{57} &= \theta \\ \Rightarrow \frac{43}{57} &= \theta + \frac{43 * \theta}{57} \\ \Rightarrow 43 &= 57 * \theta + 43 * \theta \\ \Rightarrow 43 &= 100 * \theta \\ \Rightarrow \theta &= \frac{43}{100} = 0.43 \end{aligned}$$

The difference between θ in Bernoulli distribution and Binomial distribution are 0. The binomial is, after all, the outcome of n separate Bernoulli trials.

The Bernoulli distribution, when $n = 1$, is a variant of the binomial distribution. $X \sim B(1, p)$ is equivalent to $X \sim \text{Bernoulli}(p)$ in terms of symbolism (p). Any binomial distribution, $B(n, p)$, is the sum of n separate Bernoulli trials, $\text{Bernoulli}(p)$, all of which have the same probability p [4].

References:

- [1]. Maximum likelihood method (ML), <https://www.uni-kassel.de/fb07/index.php?eID=dumpFile&t=f&f=2722&token=79679e59f57ec8195642cdfd6ad1ea6327df6f78>
- [2]. Maximum-likelihood (ML) Estimation, <https://online.stat.psu.edu/stat504/lesson/1/1.5>
- [3]. Maximum Likelihood Estimation, Example 8.8, https://www.probabilitycourse.com/chapter8/8_2_3_max_likelihood_estimation.php
- [4]. Binomial distribution, https://en.wikipedia.org/wiki/Binomial_distribution