

Maximum Likelihood Estimation

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Motivating example:

Consider a bag containing 3 balls.
Each ball is either red or blue.

Let θ be the number of blue balls
possible outcomes are $\theta = \{0, 1, 2, 3\}$.

Choose 4 balls with replacement at random.
So, let X_1, X_2, X_3, X_4 be i.i.d. r.v.

$$X_i = \begin{cases} 1 & \text{if blue} \\ 0 & \text{if red} \end{cases}$$

$$X_i \sim \text{Bernoulli}\left(\frac{\theta}{3}\right).$$

Experiment: Taking data.

observations: $x_1 = 1, x_2 = 0, x_3 = 1, x_4 = 1$.
3 blue, 1 red.

1. For each possible values of θ , what is the probability of the observed sample $(x_1, x_2, x_3, x_4) = (1, 0, 1, 1)$.

We know $\theta = \{0, 1, 2, 3\} \rightarrow$ number of blue balls.

Since $X_i \sim \text{Bernoulli}\left(\frac{\theta}{3}\right)$, then

$$P_{X_i}(x) = \begin{cases} \theta/3 & \text{if } x=1 \\ 1-\theta/3 & \text{if } x=0 \end{cases}$$

Since X_i are independent, the joint PMF for X_1, X_2, X_3 , and X_4 is

$$P_{X_1, X_2, X_3, X_4}(x_1, x_2, x_3, x_4) = P_{X_1}(x_1) P_{X_2}(x_2) P_{X_3}(x_3) P_{X_4}(x_4) = L(x_1, x_2, x_3, x_4)$$

So,

$$L(1, 0, 1, 1) = P_{X_1}(1) P_{X_2}(0) P_{X_3}(1) P_{X_4}(1)$$

$$= \left(\frac{\theta}{3}\right) \left(1 - \frac{\theta}{3}\right) \left(\frac{\theta}{3}\right) \left(\frac{\theta}{3}\right)$$

$$L(1, 0, 1, 1; \theta) = \left(\frac{\theta}{3}\right)^3 \left(1 - \frac{\theta}{3}\right)$$

$$L(1,0,1,1; \theta) = \left(\frac{\theta}{3}\right)^3 \left(1 - \frac{\theta}{3}\right)$$

The joint PMF depends on θ . $J(x_1, x_2, x_3, x_4; \theta)$.

Compute the likelihoods.

θ	$J(x_1, x_2, x_3, x_4; \theta)$	
0	0	→ because sample includes both red & blue
1	0.0247	
2	0.0988	→ maximum likelihood
3	0	→ because sample includes both red & blue

Choose $\theta=2$. So our estimate is $\hat{\theta}=2$, this means that the observed data is most likely to occur for $\theta=2$.

Likelihood function Definition

Let $x_1, x_2, x_3, \dots, x_n$ be a random sample from a distribution with parameter θ .

Suppose we observed $x_1 = x_1, x_2 = x_2, \dots, x_n = x_n$.

1. If x_i 's are discrete, then

$$L(x_1, x_2, \dots, x_n; \theta) = P_{x_1, x_2, \dots, x_n}(x_1, x_2, \dots, x_n; \theta)$$

2. If x_i 's are jointly continuous, then

$$L(x_1, x_2, \dots, x_n; \theta) = f_{x_1, x_2, \dots, x_n}(x_1, x_2, \dots, x_n; \theta)$$

In some problems, you can do log likelihood

$$\ln(L(x_1, x_2, \dots, x_n; \theta)).$$

Example:

1. Find the likelihood function of

$X_i \sim \text{Binomial}(3, \theta)$ and we have observed $(x_1, x_2, x_3, x_4) = (1, 3, 2, 2)$

$$P_{X_i}(x; \theta) = \binom{3}{x} \theta^x (1-\theta)^{3-x}$$

So,

$$\begin{aligned}L(x_1, x_2, x_3, x_4; \theta) &= P_{x_1, x_2, x_3, x_4}(x_1, x_2, x_3, x_4; \theta) \\&= P_{x_1}(x_1; \theta) P_{x_2}(x_2; \theta) P_{x_3}(x_3; \theta) P_{x_4}(x_4; \theta) \\&= \binom{3}{x_1} \binom{3}{x_2} \binom{3}{x_3} \binom{3}{x_4} \theta^{x_1+x_2+x_3+x_4} (1-\theta)^{12-(x_1+x_2+x_3+x_4)}\end{aligned}$$

plug-in observations:

$$\begin{aligned}L(1, 3, 2, 2; \theta) &= \binom{3}{1} \binom{3}{3} \binom{3}{2} \binom{3}{2} \theta^8 (1-\theta)^4 \\&= 27 \theta^8 (1-\theta)^4\end{aligned}$$

Maximum Likelihood Estimate (MLE)

Given that we observed $X_1=x_1, X_2=x_2, \dots, X_n=x_n$,
a MLE of θ , shown by $\hat{\theta}_{MLE}$ is a value of θ that
maximizes the likelihood function $L(x_1, x_2, x_3, \dots, x_n; \theta)$.

Example:

2. $X_i \sim \text{Binomial}(3, \theta)$

$$L(1, 3, 2, 2; \theta) = 27 \theta^8 (1-\theta)^4.$$

Find MLE.

$$\frac{d}{d\theta} L(1, 3, 2, 2; \theta) = 27 (8\theta^7 (1-\theta)^4 - 4\theta^8 (1-\theta)^3)$$

$$0 = 27 (8\theta^7 (1-\theta)^4 - 4\theta^8 (1-\theta)^3)$$

↓

solutions ~~$\theta=0$~~ , $\theta=2/3$, ~~$\theta=1$~~

So, $\hat{\theta}_{MLE} = \frac{2}{3}$.