

# 2.1 Method of Moment Likelihood

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## Definition:

Let  $f(X_1, \dots, X_n, \theta)$ ,  $\theta \in \Omega_\theta \subset \mathbb{R}^k$ , be the joint probability mass (or density) function of  $n$  random variables  $X_1, \dots, X_n$  with sample values  $x_1, \dots, x_n$ . The likelihood function of the sample is given by

$$L(\theta, x_1, \dots, x_n) = h(x_1, \dots, x_n), [L = (\theta), \text{ in a breifer notation}]$$

we emphasize that  $L$  is a function of  $\theta$  for fixed sample values

If  $X_1, \dots, X_n$  are discrete *i.i.d* random variables with probability mass function  $p(x, \theta)$ , then the likelihood function is given by

$$\begin{aligned} L(\theta) &= P(X_1 = x_1, \dots, X_n = x_n) \\ &= \prod_{i=1}^n P(X_i = x_i) = \prod_{i=1}^n p(x_i, \theta) \end{aligned}$$

and in a continuous case, if the probability density function is  $f(x, \theta)$ , then the likelihood function is

$$L(\theta) = \prod_{i=1}^n f(x_i, \theta)$$

## Definition:

Let  $L(\theta) = \prod_{i=1}^n p(x_i, \theta)$  (or  $L(\theta) = \prod_{i=1}^n f(x_i, \theta)$ ) be the likelihood function. Corresponding to random sample  $x_1, \dots, x_n$  drawn from the discrete *p.d.f*  $p(x, \theta)$  (or from the continuous *p.d.f*  $p(x, \theta)$ ), where  $\theta$  is an unknown parameter. Let  $\hat{\theta}$  be a value of the parameter such that  $L(\hat{\theta}) \geq L(\theta)$  for all possible values of the parameter  $\theta$ . Then  $\hat{\theta}$  is called a **maximum likelihood estimate** of  $\theta$ .

The **Maximum Likelihood Estimates** of  $\theta$  is

$$L(\hat{\theta}_l, x_1, \dots, x_n) = \max_{\theta \in \Omega_\theta} L(\theta, x_1, \dots, x_n)$$

where  $\Omega_\theta$  is the set of possible values of parameter  $\theta$

Procedure to Find Maximum Likelihood Estimate(s) (MLEs):

1. Define the likelihood function,  $L(\theta)$
2. Often it is easier to take the natural logarithm of  $L(\theta)$ ,  $\ln(L(\theta))$
3. When applicable, differentiate  $\ln(L(\theta))$  with respect to  $\theta$ , and then equate the derivative to 0
4. Solve for the parameter  $\theta$  to obtain  $\hat{\theta}$
5. Check whether it is a maximiser of global maximiser  $\hat{\theta}_l$