# Statistical Inference and Multivariate Analysis (MA324)

Lecture 12

#### Method of Moment Estimator



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Jan-May 2023

## Method of Moment Estimator (MME):

MME was first introduced by Karl Pearson in the year 1902. The basic method can be summarized in following algorithm:

- **①** Suppose that we have a RS of size n from a population with PMF/PDF  $f(x;\theta)$ , where  $\theta=(\theta_1,\ldots,\theta_k)$  is the unknown parameter vector. We want to find estimator of  $\theta$ .
- ② Calculate first k (number of unknown parameters) **moments**  $\mu'_1, \ldots, \mu'_k$  of  $f(x; \theta)$ , where  $\mu'_r = E_{\theta}(X^r)$ .
- **3** Calculate first k sample moments  $m'_1, \ldots, m'_k$ , where  $m'_r = \frac{1}{n} \sum_{i=1}^n X_i^r$ .
- **Solve** the system of k equations (if they are consistent) for  $\theta_i$ 's. The solutions are the MMEs of the unknown parameters.

### Examples:

Example 1: Let  $X_1, X_2, \ldots, X_n \overset{i.i.d.}{\sim} Bernoulli(\theta), \theta \in (0, 1) = \Theta$ . Here, we have one parameter  $\theta$ . Thus, k = 1.  $E(X_1) = \theta$ . Hence, we get the MME of  $\theta$  is  $\widehat{\theta} = \bar{X}$ .

Example 2: Let  $X_1, X_2, \ldots, X_n \stackrel{i.i.d.}{\sim} N(\mu, \sigma^2)$ ,  $\boldsymbol{\theta} = (\mu, \sigma^2) \in \mathbb{R} \times \mathbb{R}^+ = \Theta$ . Here k = 2,  $E(X) = \mu$ , and  $E(X^2) = \sigma^2 + \mu^2$ . Hence, we get the MMEs of  $\mu$  and  $\sigma^2$  are  $\widehat{\mu} = \overline{X}$  and  $\widehat{\sigma^2} = \frac{1}{n} \sum_{i=1}^n (X_i - \overline{X})^2$ , respectively.

#### Examples:

Example 3: Let  $X_1, X_2, \ldots, X_n \stackrel{i.i.d.}{\sim} N(0, \sigma^2), \sigma > 0$ . Here k = 1. However, as E(X) = 0, equating  $E(X) = \bar{X}$  does not provide any solution (inconsistent). Alternatively, we can find  $E(X^2) = \sigma^2$  and equate to  $m_2'$  to obtain MME of  $\sigma^2$  as  $\widehat{\sigma^2} = \frac{1}{n} \sum_{i=1}^n X_i^2$ .

Example 4: Let  $X_1, X_2, \ldots, X_n \overset{i.i.d.}{\sim} N(\theta, \theta^2), \theta > 0$ . Here k = 1.  $E(X) = \theta$ . Equating  $E(X) = \bar{X}$ , we get MME of  $\theta$  is  $\hat{\theta} = \bar{X}$ . However, this may not be a meaningful estimator as  $\bar{X}$  can be negative with positive probability, while  $\theta > 0$ .

Remark: Previous two examples show that there are some degrees of arbitrariness in this method.