# (Statistical Estimation & Sampling distribution)

### • Difference between Statistics and Parameter:

BASIS FOR COMPARISON	STATISTIC	PARAMETER
Meaning	Statistic is a measure which describes a fraction of population.	Parameter refers to a measure which describes population.
Numerical value	Variable and Known	Fixed and Unknown
Statistical Notation	$\stackrel{\square}{\times}$ = Sample Mean	μ = Population Mean
	s = Sample Standard Deviation	σ = Population Standard Deviation

### • Point estimate

A **point estimate** of an unknown parameter  $\theta$  is a statistic  $\hat{\theta}$  that represents a "best guess" at the value of  $\theta$ . There may be more than one sensible point estimate of a parameter.

## • Properties of a good point estimate:

### 1) Unbiased Estimates:

A point estimate  $\hat{\theta}$  for a parameter  $\theta$  is said to be **unbiased** if

$$E(\hat{\theta}) = \theta$$

Unbiasedness is a good property for a point estimate to possess. If a point estimate is not unbiased, then its bias can be defined to be

bias = 
$$E(\hat{\theta}) - \theta$$

- 2) Consistency:  $\lim_{n\to\infty} V(\hat{\theta}) = 0$
- 3) Efficiency:  $Var(\hat{\theta}) \leq Var(\theta^*)$
- 4) Sufficiency: For making inference about an unknown  $\theta$ , the statistician makes a reduction of the data by using a statistic, if no information about  $\theta$  is lost then this statistic is called sufficient estimator.

• Factorization theorem of Fisher and Neyman:

Instead of checking sufficiency via the definition it is more convenient to make use of the following theorem which gives a necessary and sufficient condition for a statistic to be sufficient.

$$f(x_1; \theta) f(x_2; \theta) \dots f(x_n; \theta) = g(t(x_1, \dots, x_n); \theta) . h(x_1, \dots, x_n)$$

where g is a nonnegative function depending on  $\theta$  and  $x_1, \ldots, x_n$  and h is a nonnegative function, not depending on  $\theta$ .

#### Classwork

 $X_1, ..., X_m$  be a random sample from a binomial distribution with parameter  $\theta$ . Find a sufficient statistic for the parameter  $\theta$ .

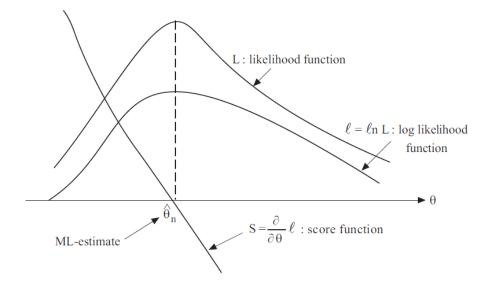
#### Homework:

- 1)  $X_1, ..., X_n$  be a random sample from a bernoulli distribution with parameter  $\theta$ . Find a sufficient statistic for the parameter  $\theta$ .
- 2)  $X_1, ..., X_n$  be a random sample from an exponential distribution with parameter  $\theta$ . Find a sufficient statistic for the parameter  $\theta$ .
- 3)  $X_1, ..., X_n$  be a random sample from a Poisson distribution with parameter  $\theta$ . Find a sufficient statistic for the parameter  $\theta$ .
- Methods of point estimate
  - 1) Maximum Likelihood Method (MLE)
  - 2) Method of moments (MM)
  - 3) Least Squares Method (LSM)
  - 1) Maximum Likelihood Method (MLE) Let  $(X_1, X_2, ..., X_n)$  be a random sample from a population distribution  $f(x; \theta)$ . The likelihood function is:

$$\mathbf{L}(\theta; X_1, ..., X_n) = \mathbf{f}(X_1; \theta) \mathbf{f}(X_2; \theta) ... \mathbf{f}(X_n; \theta) = \prod_{i=1}^{\infty} f(x_i; \theta)$$

The basic idea behind this form of the method is to:

- 1. Take logarithm of the likelihood function.
- 2. Derivate the log likelihood function with respect to parameters, and setting to 0.
- 3. Solve the equations to estimate the parameters.
- 4. To verify that we indeed obtain a maximum value of the parameters, we need to take the second order derivative of the log likelihood with respect to parameters.



The maximum likelihood principle says that, out of all the possible values for  $\theta$ , the value that maximizes the likelihood of the observed data should be chosen.

Advantage: The advantages of this method are:

- 1. consistent.
- 2. asymptotically unbiased and asymptotically the most efficient.

Disadvantage: The disadvantages of this method are:

- 1. The likelihood equations need to be specifically worked out for a given distribution.
- 2. Estimates can be heavily biased for small samples.
- 3. Can be sensitive to the choice of starting values.

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 $X_1, ..., X_m$  be a random sample from a binomial distribution with parameter  $\theta$ . Find MLE for the parameter  $\theta$ . Show that this estimator is an unbiased and consistent estimator.

#### Homework:

- 1)  $X_1, ..., X_n$  be a random sample from a bernoulli distribution with parameter  $\theta$ . Find MLE for the parameter  $\theta$ . Show that this estimator is an unbiased and consistent estimator.
- 2)  $X_1, ..., X_n$  be a random sample from an exponential distribution with parameter  $\theta$ . Find MLE for the parameter  $\theta$ . Show that this estimator is an unbiased and consistent estimator.

3)  $X_1, ..., X_n$  be a random sample from a Poisson distribution with parameter  $\theta$ . Find MLE for the parameter  $\theta$ . Show that this estimator is an unbiased and consistent estimator.