

# Method of moments & Maximum Likelihood Estimation

Tutorial-2

Course: Statistical Data Analysis

IIIT Sricity, 2020

## Estimation of parameter from Binomial Distribution

Associated to Bernouli trials

$x$  = total number of “successes” (pass or fail, heads or tails etc.)

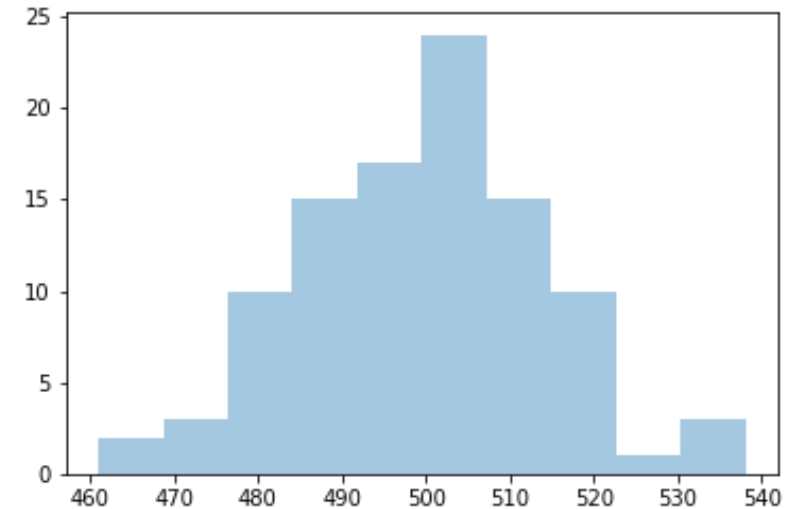
$p$  = probability of a success on an individual trial

$n$  = number of trials

- To estimate ' $\tilde{p}$ ' from Binomial Distribution

Sampled from the Binomial distribution with described parameter

- $\bar{x} = n\tilde{p}$
- sample 100 points from a binomial distribution with known parameters.



$$f(x,n,p) = {}^nC_x p^x (1-p)^{(n-x)}$$

$$\text{Mean} = np$$

$$\text{Variance} = np(1-p)$$

$$n=1000$$

$$p=0.5$$

$$\text{Mean} = 500$$

$$\text{Variance} = 250$$

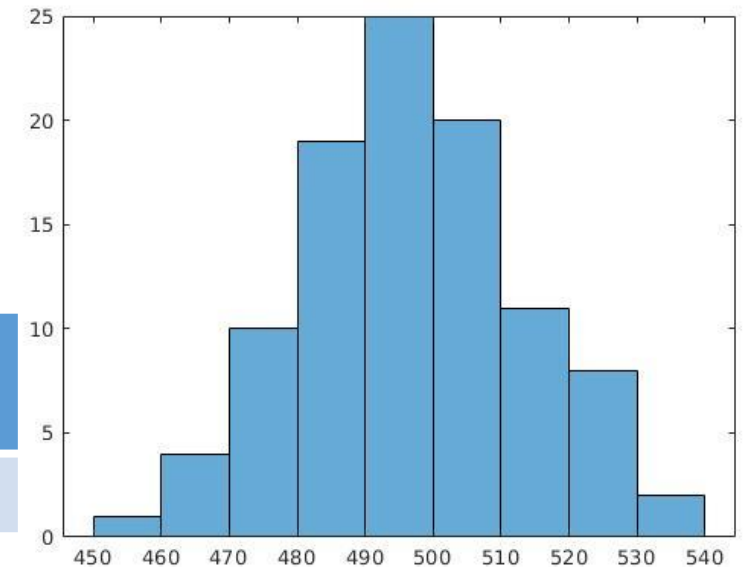
## Estimation of parameter for sample from Binomial Distribution

Compute moment of 1st order from 100 samples:

$$\bar{x} = n\tilde{p} = 496.8$$

$$\tilde{p} = (496.8/n) = 0.496$$

| Sample Size              | 100   | 200   | 300   | 400    | 500   |
|--------------------------|-------|-------|-------|--------|-------|
| Value of ' $\tilde{p}$ ' | 0.496 | 0.503 | 0.501 | 0.5007 | 0.501 |



Histogram of 100 samples

- Actual probability of a success on an individual trial= 0.5

Note:

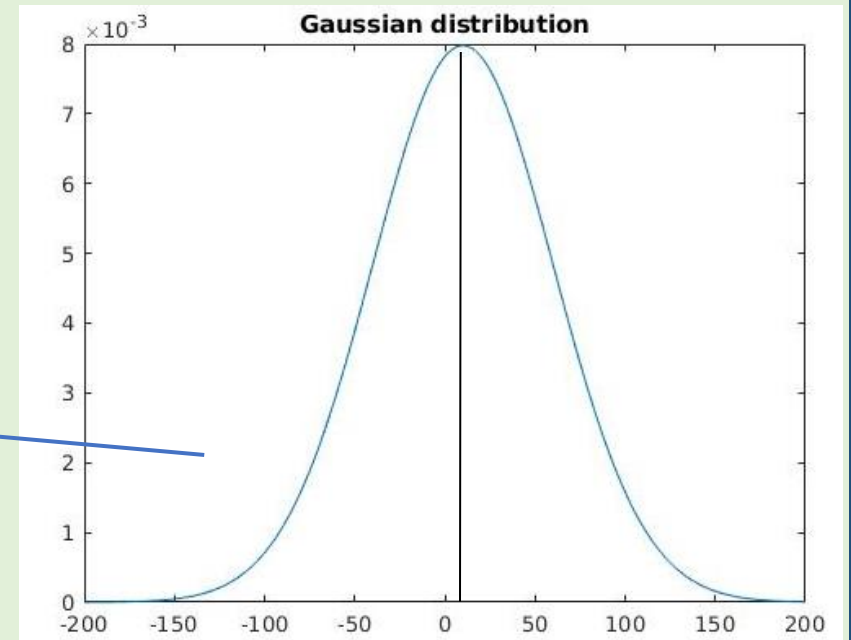
**Consistency of the estimator:**

As sample size increases the estimation becomes more accurate.

## Estimation of parameter using sample from Gaussian Distribution

- Data sampled from Gaussian distribution.
- Used 1000 samples. ←
- To be estimated Mean and variance
- Mean =  $\mu$  = 1st moment
- 1st moment = 8.36 =  $\mu$

| Sample size | 1000 | 2000  | 5000 | 10000 | 15000 |
|-------------|------|-------|------|-------|-------|
| Mean        | 8.36 | 11.64 | 9.11 | 10.22 | 9.96  |



$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2}$$

$\mu$  : mean ,  $\sigma$  standard deviation  
 $\mu = 10, \sigma = 50$

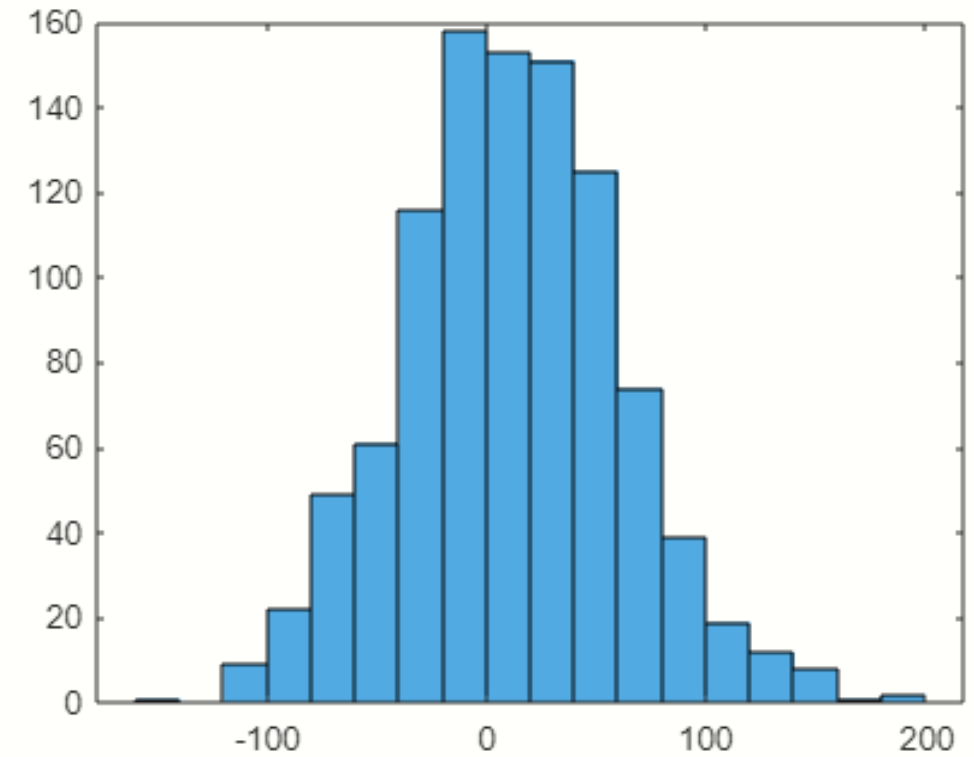
## Estimation of parameter using sample from Gaussian Distribution

For estimation of variance:

Use

- $\text{Var}(X) = E(X^2) - \{E(X)\}^2$
- Estimation for 1000 sample:  $\text{Var}(X) = 2524$

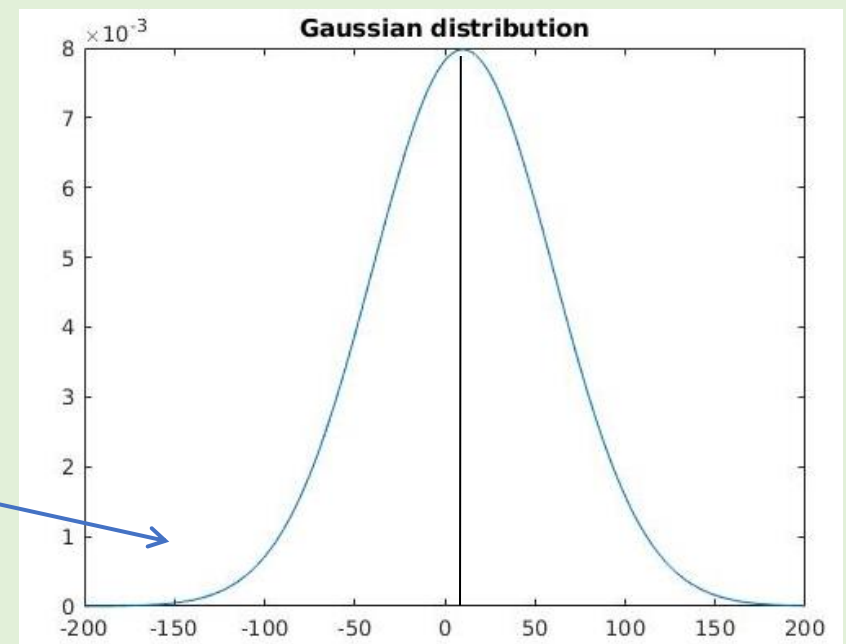
| Sample size | 1000 | 2000 | 5000 | 10000 | 15000 |
|-------------|------|------|------|-------|-------|
| Variance    | 2524 | 2500 | 2489 | 2528  | 2513  |



Histogram of 1000 samples

## Maximum Likelihood Estimation of parameter using sample from Gaussian Distribution

- Data sampled from Gaussian distribution.
- Used 1000 samples.
- To estimate Mean.
- We used s.d= 50 for the 100 samples.
- Likelihood function=  $f(x_1, x_2, \dots, x_n | \mu)$  , where  $f(x)$  given by
- Assuming the samples are from i.i.d  
We have  $f(x_1 | \mu) f(x_2 | \mu) \dots f(x_n | \mu)$

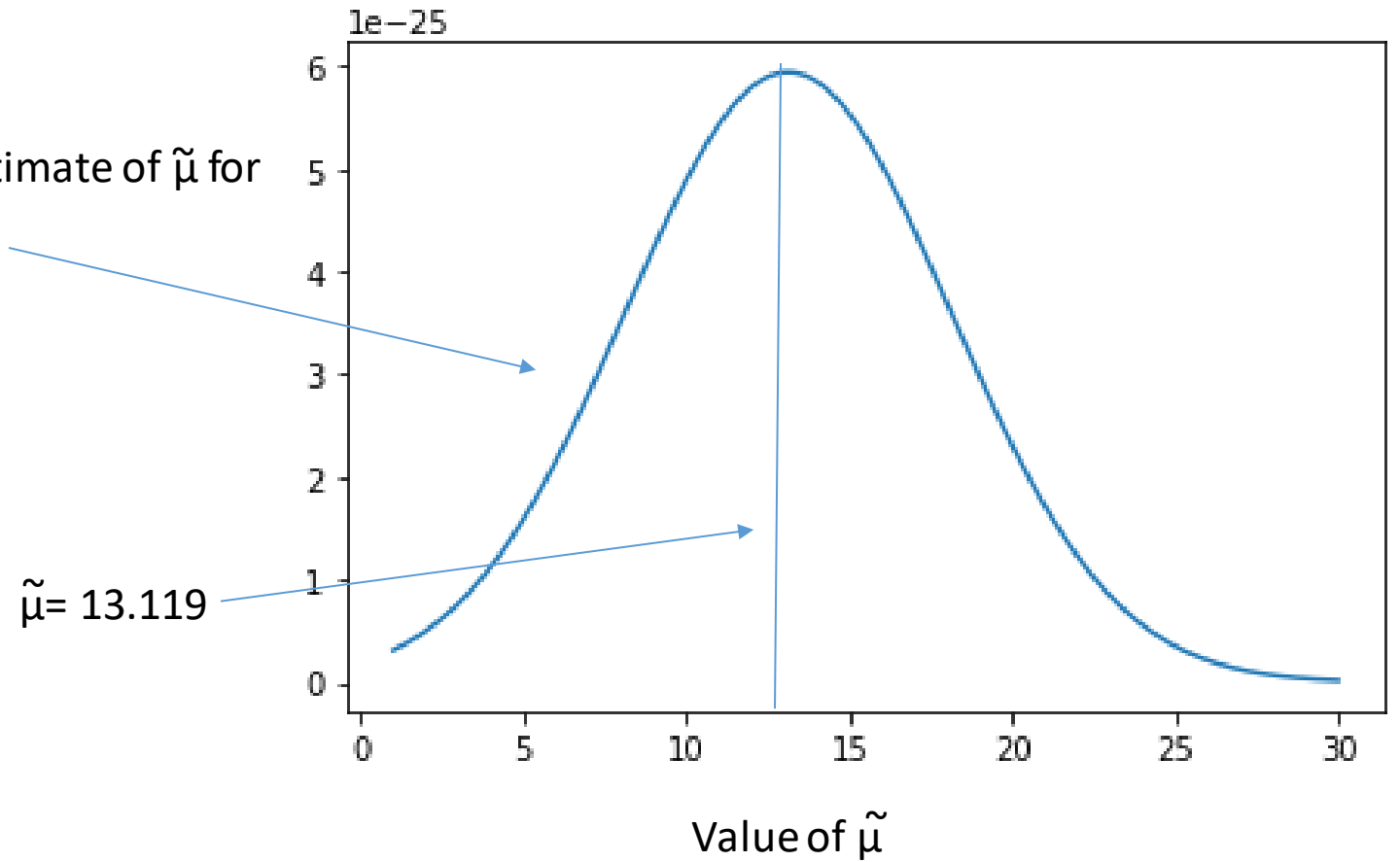


$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2}$$

$\mu$  :mean ,  $\sigma$  standard deviation  
 $\mu = 10, \sigma = 50$

## Maximum Likelihood Estimation of parameter using sample from Gaussian Distribution

- Sample mean=13.12      Likelihood estimate of  $\tilde{\mu}$  for 100 samples
- $\tilde{\mu} = 13.119$ , Likelihood estimation gives maximum value



- To estimate population variance ( $\sigma^2$ )

For the 100 samples,  
taken,  $\mu=10$  (known)

Likelihood estimate  
of  $\sigma^2$  for 100 samples

$$\frac{1}{n} \sum_{i=1}^n (x_i - \mu)^2 = 2501.66$$

- Likelihood function maximizes at variance value 2501.6.

2501.6

