

STAT 2593

Lecture 024 - Methods of Point Estimation

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Methods of Point Estimation

Learning Objectives

1. Understand the method of moments estimation procedure.



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- ▶ Estimation techniques are designed to generate estimators for parameters of interest, generally.
 - ▶ **Maximum likelihood estimation** is the most prominent estimation technique.
 - ▶ **Method of moments** is an intuitive technique, which typically generates robust estimators.

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 - ▶ In general, taking $E[X^k]$ is called the k -th moment.
- ▶ Population moments will be functions of the unknown parameters, denoted θ .

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$$\widehat{m}_k = \frac{1}{n} \sum_{i=1}^n X_i^k.$$

- ▶ With the estimator \widehat{m}_k , we can use a sample to compute estimates (i.e., values) for these.

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 3. We set $\hat{m}_k = E[X^k]$, for $k = 1, \dots, L$, and then solve for the components of Θ .
 4. The solution to the system of equations gives $\hat{\Theta} = (\hat{\theta}_1, \dots, \hat{\theta}_L)$.

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- ▶ The method of moments estimators will generally be **biased**.
- ▶ Method of moment estimators do not need to be **valid** parameter values.

Summary

- ▶ Estimation techniques are general processes which generate estimators.
- ▶ The method of moments estimators are derived by setting population moments equal to sample moments, and solving for the unknown parameters.
- ▶ Method of moments estimators are robust and intuitive, but generally biased, and may not exist (or be valid).