AUA CS108, Statistics, Fall 2020 Lecture 32

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Contents

► Method of Moments (MoM, MME)

Methods to find (good) Estimators

The Problem

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We are given a Random Sample from a Parametric Family of Distributions,

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Problem: The Problem is to find/construct a good Estimator for θ , using our Random Sample.

The Method of Moments

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Note: Note that, in general, the Theoretical Moments of \mathcal{F}_{θ} are functions of θ .

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Note: The Empirical Moment is independent of the Parameter θ , it is just a Statistics, it is a function of $X_1, X_2, ..., X_n$, it is a r.v.

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from the following Model:

$$\begin{array}{c|c|c} X & 0 & 1 & 2 \\ \hline \mathbb{P}(X=x) & \frac{\theta}{10} & \frac{\theta}{5} & 1 - \frac{3\theta}{10}, \end{array}$$

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Example: Find the MoM Estimator for λ in the $Exp(\lambda)$ Model.

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