# MATH 450: Mathematical statistics 

September 24th, 2020

Lecture 8: Method of moments

| Week $2 \ldots \ldots$. | Chapter 6: Statistics and Sampling <br> Distributions |
| :--- | :--- |
| Week $4 \ldots \ldots$. | Chapter 7: Point Estimation |
| Week $6 \ldots \ldots$. | Chapter 8: Confidence Intervals |
| Week $9 \ldots \ldots$. | Chapter 9: Test of Hypothesis |
| Week $11 \ldots \ldots$. | Chapter 10: Two-sample inference |
| Week $12 \ldots \ldots$. | Regression |

## Chapter 7: Overview

7.1 Point estimate

- unbiased estimator
- mean squared error
7.2 Methods of point estimation
- method of moments
- method of maximum likelihood.
7.3 Sufficient statistic
7.4 Information and Efficiency
- Large sample properties of the maximum likelihood estimator


## Question of this chapter

- Given a random sample $X_{1}, \ldots, X_{n}$ from a distribution with pmf/pdf $f(x, \theta)$ parameterized by a parameter $\theta$
- Goal: Estimate $\theta$



## Point estimate

$$
f(x, \theta)
$$



## Definition

A point estimate $\hat{\theta}$ of a parameter $\theta$ is a single number that can be regarded as a sensible value for $\theta$.

$$
\begin{aligned}
\text { population parameter } & \Longrightarrow \text { sample } \\
\theta & \Longrightarrow X_{1}, X_{2}, \ldots, X_{n}
\end{aligned}
$$

## Estimate vs estimator

$$
\begin{aligned}
\text { sample } & \Longrightarrow \text { estimator } \\
X_{1}, X_{2}, \ldots, X_{n} & \Longrightarrow \hat{\theta} \\
\text { observed data } & \Longrightarrow \text { estimate } \\
x_{1}, x_{2}, \ldots, x_{n} & \Longrightarrow \hat{\theta}
\end{aligned}
$$

## Mean Squared Error

- Measuring error of estimation

$$
|\hat{\theta}-\theta| \quad \text { or } \quad(\hat{\theta}-\theta)^{2}
$$

- The error of estimation is random


## Definition

The mean squared error of an estimator $\hat{\theta}$ is

$$
E\left[(\hat{\theta}-\theta)^{2}\right]
$$

## Bias-variance decomposition

## Theorem

$$
\operatorname{MSE}(\hat{\theta})=E\left[(\hat{\theta}-\theta)^{2}\right]=V(\hat{\theta})+(E(\hat{\theta})-\theta)^{2}
$$

Bias-variance decomposition
Mean squared error $=$ variance of estimator $+(\text { bias })^{2}$

## Bias-variance decomposition

Low bias
High bias


## Statistical bias vs. social bias

How things should be


## Unbiased estimators

## Definition

A point estimator $\hat{\theta}$ is said to be an unbiased estimator of $\theta$ if

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

for every possible value of $\theta$.

Unbiased estimator
$\Leftrightarrow$ Bias $=0$
$\Leftrightarrow$ Mean squared error $=$ variance of estimator

## Example: sample proportion

## Problem

Let $X_{1}, X_{2}, \ldots, X_{n}$ be a random sample of size $n$ from a Bernoulli distribution with probability of success $p$

| $x$ | 0 | 1 |
| :---: | :---: | :---: |
| $p(x)$ | $1-p$ | $p$ |

Assume that we estimate $p$ by using the sample mean

$$
\bar{X}=\frac{X_{1}+X_{2}+\ldots X_{n}}{n}
$$

What are the bias and the variance of this estimator?

## Example: sample proportion

## Problem

Let $X_{1}, X_{2}, \ldots, X_{n}$ be a random sample of size $n$ from a Bernoulli distribution with probability of success $p$

| $x$ | 0 | 1 |
| :---: | :---: | :---: |
| $p(x)$ | $1-p$ | $p$ |

Assume that we estimate $p$ by using the sample mean

$$
\bar{X}=\frac{X_{1}+X_{2}+\ldots X_{n}}{n}
$$

Compute the MSE of this estimator.

## Example: sample proportion

## Problem

Let $X_{1}, X_{2}, \ldots, X_{n}$ be a random sample of size $n$ from a Bernoulli distribution with probability of success $p$

| $x$ | 0 | 1 |
| :---: | :---: | :---: |
| $p(x)$ | $1-p$ | $p$ |

Assume that we estimate $p$ by using

$$
\tilde{p}=\frac{X_{1}+X_{2}+\ldots+X_{n}+2}{n+4}
$$

Compute the MSE of this estimator.

## Example 7.1 and 7.4




## Bias-variance decomposition

Low bias
High bias


## Method of moments

## Example

## Problem

Let $X_{1}, \ldots, X_{10}$ be a random sample from a distribution with pdf

$$
f(x)=\left\{\begin{array}{l}
(\theta+1) x^{\theta} \quad \text { if } 0 \leq x \leq 1 \\
0 \quad \text { otherwise }
\end{array}\right.
$$

A random sample of ten students yields data

$$
\begin{aligned}
& x_{1}=.92, x_{2}=.79, x_{3}=.90, x_{4}=.65, x_{5}=.86, \\
& x_{6}=.47, x_{7}=.73, x_{8}=.97, x_{9}=.94, x_{10}=.77
\end{aligned}
$$

Provide an estimate of $\theta$.

- We can compute $E(X) \rightarrow$ the answer will be a function of $\theta$
- For large $n$, we have

$$
\bar{X}=\frac{X_{1}+X_{2}+\ldots+X_{n}}{n}
$$

is close to $E[X]$

- We can compute $\bar{x}$ from the data $\rightarrow$ approximate $\theta$


## Example 1: Step 1

## Problem

Let $X_{1}, \ldots, X_{10}$ be a random sample from a distribution with pdf

$$
f(x)=\left\{\begin{array}{l}
(\theta+1) x^{\theta} \quad \text { if } 0 \leq x \leq 1 \\
0 \quad \text { otherwise }
\end{array}\right.
$$

Compute $E[X]$.

## Example 1

## Problem

A random sample of ten students yields data

$$
\begin{aligned}
& x_{1}=.92, x_{2}=.79, x_{3}=.90, x_{4}=.65, x_{5}=.86, \\
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\end{aligned}
$$

Compute $\bar{x}$.

## Example 1

## Problem

Let $X_{1}, \ldots, X_{10}$ be a random sample from a distribution with pdf

$$
f(x)=\left\{\begin{array}{l}
(\theta+1) x^{\theta} \quad \text { if } 0 \leq x \leq 1 \\
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\end{aligned}
$$

Provide an estimate of $\theta$ by the method of moments.

## Method of moments: Example 2

## Problem

Suppose that for a parameter $0 \leq \theta \leq 1, X$ is the outcome of the roll of a four-sided tetrahedral die

| $x$ | 1 | 2 | 3 | 4 |
| :---: | :---: | :---: | :---: | :---: |
| $p(x)$ | $\frac{3 \theta}{4}$ | $\frac{\theta}{4}$ | $\frac{3(1-\theta)}{4}$ | $\frac{(1-\theta)}{4}$ |

Suppose the die is rolled 10 times with outcomes

$$
4,1,2,3,1,2,3,4,2,3
$$

Use the method of moments to obtain an estimate of $\theta$.

## Moments

- Let $X_{1}, \ldots, X_{n}$ be a random sample from a normal distribution with pmf or pdf $f(x)$.
- For $k=1,2,3, \ldots$, the $k^{\text {th }}$ population moment, or $k^{t h}$ moment of the distribution $f(x)$, is

$$
E\left(X^{k}\right)
$$

- First moment: the mean
- Second moment: $E\left(X^{2}\right)$


## Sample moments

- Let $X_{1}, \ldots, X_{n}$ be a random sample from a distribution with pmf or pdf $f(x)$.
- For $k=1,2,3, \ldots$, the $k^{\text {th }}$ sample moment is

$$
\frac{X_{1}^{k}+X_{2}^{k}+\ldots+X_{n}^{k}}{n}
$$

The law of large numbers provides that when $n \rightarrow \infty$

$$
\frac{X_{1}^{k}+X_{2}^{k}+\ldots+X_{n}^{k}}{n} \rightarrow E\left(X^{k}\right)
$$

## Method of moments: ideas

- Let $X_{1}, \ldots, X_{n}$ be a random sample from a distribution with pmf or pdf

$$
f\left(x ; \theta_{1}, \theta_{2}, \ldots, \theta_{m}\right)
$$

- Assume that for $k=1, \ldots, m$

$$
\hat{u}_{k}=\frac{X_{1}^{k}+X_{2}^{k}+\ldots+X_{n}^{k}}{n}=E\left(X^{k}\right)
$$

- Solve the system of equations for $\theta_{1}, \theta_{2}, \ldots, \theta_{m}$


## Method of moments: Example 3

## Problem

Let $\beta>1$ and $X_{1}, \ldots, X_{n}$ be a random sample from a distribution with $p d f$

$$
f(x)= \begin{cases}\frac{\beta}{x^{\beta+1}} & \text { if } x>1 \\ 0 & \text { otherwise }\end{cases}
$$

Use the method of moments to obtain an estimator of $\beta$.

## Method of moments: Example 4

## Problem

Let $X_{1}, \ldots, X_{n}$ be a random sample from the normal distribution $\mathcal{N}\left(0, \sigma^{2}\right)$.
Use the method of moments to obtain an estimator of $\sigma$.

