## Point Estimation

Chapter 10, Miller and Miller

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## Big Concepts of the Day

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■ Last week's Grand Question: What do sample statistics tell us about population parameters?
■ Today's Grand Question: Which sample statistics tell us about which population parameters?

## What is Point Estimation?

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Examples: $\bar{X}$ is a point estimator of $\mu . S^{2}$ is a point estimator of $\sigma^{2}$

## Desirable properties of Estimators

- Unbiasedness
- Minimum Variance or Efficiency
- Consistency


## Unbiased Estimator

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Last class we showed $E(\bar{X})=\mu$, so $\bar{X}$ is an unbiased estimator of $\theta$.

## Unbiasedness - Intuition



Biased and unbiased estimators.

## Example 1

If $X$ is a binomial distribution with the parameters $n$ and $\theta$, show that the sample proportion, $X / n$ is an unbiased estimator of $\theta$.

## Example 2

Given a random sample of size $n$ from a population that has known mean $\mu$ and finite variance $\sigma^{2}$, show that

$$
\frac{1}{n} \sum_{i=1}^{n}\left(X_{i}-\mu\right)^{2}
$$

is an unbiased estimator of $\sigma^{2}$.

## Example 3

If $X$ is a binomial distribution with the parameters $n$ and $\theta$. Is $(X+1) /(n+2)$ an unbiased estimator of $\theta$ ?

## Asymptotically Unbiased Estimator

## Definition

Let $b_{n}(\theta)=E(\hat{\Theta})-\theta$ express the bias of an estimator $\hat{\Theta}$ based on a random sample of size $n$ from a given distribution. We say $\hat{\theta}$ is an asymptotically unbiased estimator of $\theta$ if and only if

$$
\lim _{n \rightarrow \infty} b_{n}(\theta)=0
$$

## Example 3 - Again

If $X$ is a binomial distribution with the parameters $n$ and $\theta$. Is $(X+1) /(n+2)$ an asymptotically unbiased estimator of $\theta$ ?

## Mystery of Sample Variance

If $S^{2}$ is the variance of a random sample from an infinite population with the finite variance $\sigma^{2}$ then $E\left(S^{2}\right)=\sigma^{2}$.

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## Efficiency

## Definition

If the estimate $\hat{\Theta}$ for the parameter $\theta$ of a given distribution that has the smallest variance of all unbiased estimators for $\theta$ is called the minimum variance unbiased estimator or the best unbiased estimator for $\theta$.

## Efficiency - Intuition



Distribution of three estimators of $\theta$.

## Consistency

## Definition

$\hat{\Theta}$ is said to be a consistent estimator of $\theta$ if it approaches the true value $\theta$ as the sample size gets larger and larger. Formally

$$
\lim _{n \rightarrow \infty} P(|\hat{\Theta}-\theta|<\delta)=1 \quad \forall \delta>0
$$

## Consistency - Intuition



The distribution of $\hat{\theta}$ as sample size increases.

## Possible Complications



Tradeoff between bias and variance.

## Ways to get (desirable) Estimators

■ Method of Moments

- Method of Maximum Likelihood
- Bayesian estimation


## Method of Moments

■ The kth sample moment of a set of observations $x_{1}, x_{2}, \ldots$, $x_{n}$ is the mean of their $k$ th powers, and is denoted by $m_{k}^{\prime}$

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- If a population has $r$ parameters then the method of moments consists of solving the system of equations

$$
m_{k}^{\prime}=\mu_{k}^{\prime} \quad k=1,2, \ldots, r
$$

## Example 4

Given a random sample of size $n$ from a Poisson population, use the method of moments to obtain an estimator for the parameter $\lambda$.

## Method of Maximum Likelihood

If $x_{1}, x_{2}, \ldots, x_{n}$ are values of a random sample from a population with a parameter $\theta$, the likelihood function is given by

$$
L(\theta)=P\left(X_{1}=x_{1}, X_{2}=x_{2}, \ldots, X_{n}=x_{n}\right)=f\left(x_{1}, x_{2}, \ldots, x_{n} ; \theta\right)
$$

for values of $\theta$ within a given domain. The value of $\theta$ which maximized $L(\theta)$ is called the maximum likelihood estimator of $\theta$.

## Example 4 - Again

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## Bayesian Estimation

The difference between frequentist statisticians and Bayesian statisticians has to do with whether a statistician thinks of a parameter as some unknown constant or as a random variable. Let's take a look at a simple example in an attempt to emphasize the difference.

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- Priors and Posteriors!

