Mathematical statistics

March 15^{th}, 2018

Lecture 14: Method of maximum likelihood

Where are we?

Week 1	Probability reviews
Week 2	Chapter 6: Statistics and Sampling Distributions
Week 4	**Chapter 7: Point Estimation**
Week 7	Chapter 8: Confidence Intervals
Week 10	Chapter 9: Test of Hypothesis
Week 14	Regression

Overview

7.1 Point estimate
- unbiased estimator
- mean squared error

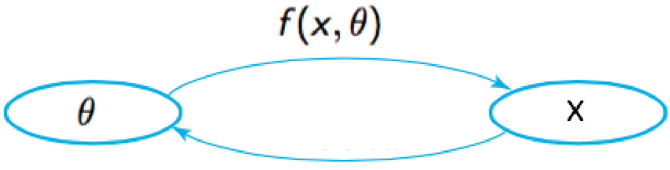
7.2 Methods of point estimation
- method of moments
- method of maximum likelihood.

7.3 Sufficient statistic

7.4 Information and Efficiency
- Large sample properties of the maximum likelihood estimator

- Bootstrap

Point estimate



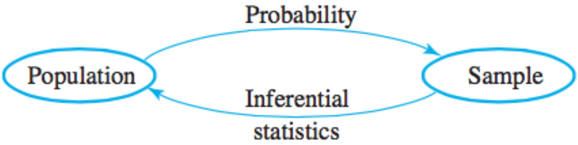
Definition

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

$$\begin{array}{ccccc} \text{population parameter} & \Longrightarrow & \textit{sample} & \Longrightarrow & \textit{estimate} \\ \theta & \Longrightarrow & X_1, X_2, \ldots, X_n & \Longrightarrow & \hat{\theta} \end{array}$$

Method of maximum likelihood

Random sample



Definition

The random variables $X_1, X_2, ..., X_n$ are said to form a (simple) random sample of size n if

1. the X_i's are independent random variables
2. every X_i has the same probability distribution

Independent random variables

Definition

Two random variables X and Y are said to be independent if for every pair of x and y values,

$$P(X = x, Y = y) = P_X(x) \cdot P_Y(y) \quad \text{if the variables are discrete}$$

or

$$f(x, y) = f_X(x) \cdot f_Y(y) \quad \text{if the variables are continuous}$$

Random sample

Let $X_1, X_2, ..., X_n$ be a random sample of size n from a distribution with density function $f_X(x)$.

Then the density of the joint distribution of $(X_1, X_2, ..., X_n)$ is

$$f_{joint}(x_1, x_2, \ldots, x_n) = \prod_{i=1}^{n} f_X(x_i)$$

Maximum likelihood estimator

- Let $X_1, X_2, ..., X_n$ have joint pmf or pdf

$$f_{joint}(x_1, x_2, \ldots, x_n; \theta)$$

where θ is unknown.

- When $x_1, \ldots, x_n$ are the observed sample values and this expression is regarded as a function of θ, it is called the likelihood function.
- The maximum likelihood estimates θ_{ML} are the value for θ that maximize the likelihood function:

$$f_{joint}(x_1, x_2, \ldots, x_n; \theta_{ML}) \geq f_{joint}(x_1, x_2, \ldots, x_n; \theta) \qquad \forall\theta$$

How to find the MLE?

- Step 1: Write down the likelihood function.
- Step 2: Can you find the maximum of this function?
- Step 3: Try taking the logarithm of this function.
- Step 4: Find the maximum of this new function.

To find the maximum of a function of θ:

- compute the derivative of the function with respect to θ
- set this expression of the derivative to 0
- solve the equation

Example

Problem

Let $X_1, \ldots, X_{10}$ be a random sample from the exponential distribution with parameter λ, that is

$$f(x; \lambda) = \lambda e^{-\lambda x}, \qquad x \geq 0$$

The observed data are

$$3.11, 0.64, 2.55, 2.20, 5.44,$$

$$3.42, 1.39, 8.13, 1.82, 1.30$$

Use the method of maximum likelihood to obtain an estimator of λ.

Example

Problem

Let $X_1, \ldots, X_{10}$ be a random sample from a distribution with pdf

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

A random sample of ten students yields data

$$x_1 = .92,\ x_2 = .79,\ x_3 = .90,\ x_4 = .65,\ x_5 = .86,$$

$$x_6 = .47,\ x_7 = .73,\ x_8 = .97,\ x_9 = .94,\ x_{10} = .77$$

Use the method of maximum likelihood to obtain an estimator of θ.

Example

Problem

Let $X_1, \ldots, X_n$ *be a random sample from the normal distribution* $\mathcal{N}(0, \sigma^2)$, *that is*

$$f(x, \theta) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{x^2}{2\sigma^2}}$$

Use the method of maximum likelihood to obtain an estimator of σ.

Example

Problem

Let $\beta > 1$ *and* $X_1, \ldots, X_n$ *be a random sample from a distribution with pdf*

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

Use the method of maximum likelihood to obtain an estimator of β.