

MITx: 14.310x Data Analysis for Social Scientists

Heli



Bookmarks

- Module 1: The Basics of R and Introduction to the Course
- ▶ Entrance Survey
- Module 2: Fundamentals of Probability, Random Variables, Distributions, and Joint Distributions
- Module 3: Gathering and Collecting Data, Ethics, and Kernel Density Estimates
- Module 4: Joint,
 Marginal, and
 Conditional
 Distributions &
 Functions of Random
 Variable

Module 7: Assessing and Deriving Estimators - Confidence Intervals, and Hypothesis Testing > Assessing and Deriving Estimators > Examples of Maximum Likelihood Estimation, Part I - Quiz

Examples of Maximum Likelihood Estimation, Part I - Quiz

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Question 1

1.0/1.0 point (graded)

For an i.i.d. random variable, the likelihood function is simply equal to: (Select all that apply)

- a. The joint PDF of the data
- lacksquare b. $\Pi_i f(heta|x)$
- lacksquare c. $\Pi_i f(x| heta)$
- lacksquare d. f(x| heta)
- lacksquare e. f(heta|x)



Explanation

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- Module 5: Moments of a Random Variable,
 Applications to Auctions,
 Intro to Regression
- Module 6: Special
 <u>Distributions, the</u>

 <u>Sample Mean, the</u>
 <u>Central Limit Theorem,</u>
 and Estimation
- Module 7: Assessing and Deriving Estimators - Confidence Intervals, and Hypothesis Testing

<u>Assessing and Deriving</u> Estimators

Finger Exercises due Nov 14, 2016 at 05:00 IST

<u>Confidence Intervals and</u> <u>Hypothesis Testing</u>

Finger Exercises due Nov 14, 2016 at 05:00 IST

Module 7: Homework

<u>Homework due Nov 07, 2016 at 05:00 IST</u>

The likelihood function is a reinterpretation of the joint PDF of our data or random sample. The likelihood function is the same as the joint PDF of the data, which for i.i.d. random variables is equal to $\Pi_i f(x|\theta)$

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You have used 1 of 2 attempts

Question 2

1.0/1.0 point (graded)

Sometimes the likelihood function is computationally difficult to maximize. In this case, what could we maximize instead? (Select all that apply.)

- a. The log of the likelihood function.
- b. The inverse of the likelihood function.
- c. The sine of the likelihood function.
- d. Any monotonic transformation of the likelihood function.



Explanation

Maximizing any monotonic transformation of the likelihood function will give us the same parameters as maximizing the original likelihood function. Taking the log is a monotonic transformation, so (a) is correct. In practice, we often maximize the log-likelihood rather than the likelihood because it is often

Exit Survey

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computationally easier to maximize the log-likelihood.

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Discussion

Topic: Module 7 / Examples of Maximum Likelihood Estimation, Part I -Quiz

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