

7. Estimating the Parameter for an Exponential Model

Estimating the Parameter for an Exponential Model

[Start of transcript. Skip to the end.](#)

Estimator

▶ Density of T_1 :

$$f(t) = \lambda e^{-\lambda t}, \quad \forall t \geq 0.$$

▶ $\mathbb{E}[T_1] = \frac{1}{\lambda}.$

▶ Hence, a natural esti

$$\frac{1}{n} \sum_{i=1}^n T_i.$$

▶ A natural estimator of λ is

(Caption will be displayed when you start playing the video.)

OK, so as I said, the density is lambda e to the minus lambda T.

You can probably convince yourself just by looking at it that this thing will integrate to one on zero infinity.

I won't do it, but you can check that the expectation is one over lambda.

That's probably a pretty standard exercise.

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Consistency and Biasedness

3/4 points (graded)

Let $X_1, X_2, \dots, X_n \stackrel{iid}{\sim} \exp(\lambda)$. Let $\bar{X}_n := \frac{1}{n} \sum_{i=1}^n X_i$ denote the sample mean of the data set.

To which value does \bar{X}_n converge (both a.s. and in probability) as $n \rightarrow \infty$?
(Choose all that apply)

☒ $\mathbb{E}[X_i]$ ✓

☐ $\frac{1}{\mathbb{E}[X_i]}$

☐ $\mathbb{E}\left[\frac{1}{X_i}\right]$

☐ λ

☒ $\frac{1}{\lambda}$ ✓

✓

To which value does $\frac{1}{\bar{X}_n}$ converge (both a.s. and in probability) as $n \rightarrow \infty$? (Choose all that apply)

☐ $\mathbb{E}[X_i]$

☒ $\frac{1}{\mathbb{E}[X_i]}$ ✓

☒ $\mathbb{E}\left[\frac{1}{X_i}\right]$

☒ λ ✓

☐ $\frac{1}{\lambda}$

✗

Which of the following is the bias of $\frac{1}{\bar{X}_n}$ as an estimator of λ ? (Choose all that apply.)

☒ $\mathbb{E}\left[\frac{1}{\bar{X}_n}\right] - \lambda$ ✓

☒ $\mathbb{E}\left[\frac{1}{\bar{X}_n}\right] - \frac{1}{\mathbb{E}[X_i]}$ ✓

☒ $\mathbb{E}\left[\frac{1}{\bar{X}_n}\right] - \frac{1}{\mathbb{E}[\bar{X}_n]}$ ✓

☐ $\frac{1}{\mathbb{E}[X_i]} - \lambda$

☐ $\frac{1}{\mathbb{E}[X_i]} - \frac{1}{\mathbb{E}[\bar{X}_n]}$

✓

Which of the following are properties of $\frac{1}{\bar{X}_n}$ as an estimator of λ ? (Choose all that apply.)

☒ consistent ✓

☐ unbiased

✓

Solution:

- By the (strong/weak) law of large numbers

$$\bar{X}_n = \frac{\sum_{i=1}^n X_i}{n} \xrightarrow[n \rightarrow \infty]{a.s./\mathbf{P}} \mathbb{E}[X_i] = \frac{1}{\lambda}.$$

- On the other hand, by the continuous mapping theorem

$$\frac{1}{\bar{X}_n} \xrightarrow[n \rightarrow \infty]{a.s./\mathbf{P}} \frac{1}{\mathbb{E}[X_i]} = \lambda.$$

- Hence, we can answer the last part immediately: $\frac{1}{\overline{X}_n}$ is a consistent estimator of λ .
- However,

$$\mathbb{E}\left[\frac{1}{\overline{X}_n}\right] \neq \frac{1}{\mathbb{E}\left[\overline{X}_n\right]} = \lambda.$$

So the bias of $\frac{1}{\overline{X}_n}$ as an estimator of $\lambda = \frac{1}{\mathbb{E}\left[X_i\right]} = \frac{1}{\mathbb{E}\left[\overline{X}_n\right]}$ is

$$\text{Bias} = \mathbb{E}\left[\frac{1}{\overline{X}_n}\right] - \frac{1}{\mathbb{E}\left[\overline{X}_n\right]}.$$

Remark: Since the function $\frac{1}{x}$ is convex (by the shape of its graph or by $\left(\frac{1}{x}\right)'' = \frac{2}{x^3} > 0$), Jensen's inequality gives

$\mathbb{E}\left[\frac{1}{\overline{X}_n}\right] > \frac{1}{\mathbb{E}\left[\overline{X}_n\right]}$ and hence the bias is greater than zero.

提交

你已经尝试了2次（总共可以尝试2次）

i Answers are displayed within the problem

Review: Central Limit Theorem

1/1 point (graded)
 The **Central Limit Theorem** states that if X_1, \dots, X_n are i.i.d. and

$$\mathbb{E}\left[X_1\right] = \mu < \infty \quad ; \quad \text{Var}\left(X_1\right) = \sigma^2 < \infty,$$

then

$$\sqrt{n}\left[\left(\frac{1}{n}\sum_{i=1}^nX_i\right) - \mu\right] \xrightarrow[n \rightarrow \infty]{(d)} Z \quad \text{ where } Z \sim \mathcal{N}(0, ?).$$

What is $\text{Var}\left(Z\right)$? (Express your answer in terms of n, μ and σ).

Var (Z) =

sigma^2

✓

Answer: sigma^2

σ²

STANDARD NOTATION

Solution:

For any n ,

$$\text{Var}\sqrt{n}\left(\overline{X}_n - \mu\right) = n\text{Var}\left(\overline{X}_n\right) = \text{Var}\left(X_i\right) = \sigma^2.$$

The central limit theorem states as $n \rightarrow \infty$, the distribution of $\sqrt{n}\left(\overline{X}_n - \mu\right)$ becomes Gaussian with the variance above (and mean 0); that is,

$$\sqrt{n}\left(\overline{X}_n - \mu\right) \xrightarrow[n \rightarrow \infty]{(d)} \mathcal{N}\left(0, \sigma^2\right).$$

Note: The variance of Z is called the **asymptotic variance** of \overline{X}_n , even though it equals the variance of $\sqrt{n}\overline{X}_n$.