

3. The sample mean

Problem 3. The sample mean

5/5 points (graded)

Let \mathbf{X} be a continuous random variable. We know that it takes values between **0** and **6**, but we do not know its distribution or its mean and variance, although we know that its variance is at most **4**. We are interested in estimating the mean of \mathbf{X} , which we denote by \mathbf{h} . To estimate \mathbf{h} , we take \mathbf{n} i.i.d. samples $\mathbf{X}_1, \dots, \mathbf{X}_n$, which all have the same distribution as \mathbf{X} , and compute the sample mean

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

1. Express your answers for this part in terms of \mathbf{h} and \mathbf{n} using standard notation.

$\mathbf{E}[H] =$

✓ Answer: h

Given the available information, the smallest upper bound for $\mathbf{Var}(H)$ that we can assert/guarantee is:

$\mathbf{Var}(H) \leq$

✓ Answer: 4/n

2. Calculate the smallest possible value of \mathbf{n} such that the standard deviation of \mathbf{H} is guaranteed to be at most 0.01.

This minimum value of \mathbf{n} is:

✓ Answer: 40000

3. We would like to be at least **96%** sure that our estimate is within **0.02** of the true mean \mathbf{h} . Using the Chebyshev inequality, calculate the minimum value of \mathbf{n} that will achieve this.

This minimum value of \mathbf{n} is:

✓ Answer: 250000

4. Suppose now that \mathbf{X} is uniformly distributed on $[\mathbf{h} - \mathbf{3}, \mathbf{h} + \mathbf{3}]$, for some unknown \mathbf{h} . Using the Central Limit Theorem, identify the most appropriate expression for a **95%** confidence interval for \mathbf{h} . You may want to refer to the normal table.

Normal Table

Show

☐ $\left[H - \frac{\sqrt{1.96 \cdot 3}}{\sqrt{n}}, H + \frac{\sqrt{1.96 \cdot 3}}{\sqrt{n}} \right]$

☐ $\left[H - \frac{1.96}{\sqrt{3n}}, H + \frac{1.96}{\sqrt{3n}} \right]$

☒ $\left[H - \frac{1.96 \cdot \sqrt{3}}{\sqrt{n}}, H + \frac{1.96 \cdot \sqrt{3}}{\sqrt{n}} \right]$ ✓

$$\bullet \left[H - \frac{1.96 \cdot 3}{\sqrt{n}}, H + \frac{1.96 \cdot 3}{\sqrt{n}} \right]$$

STANDARD NOTATION

Solution:

1. We have

$$\begin{aligned} H &= \frac{X_1 + X_2 + \dots + X_n}{n}, \\ \mathbf{E}[H] &= \frac{\mathbf{E}[X_1 + \dots + X_n]}{n} = \frac{n \cdot \mathbf{E}[X]}{n} = h, \\ \sigma_H^2 &= \text{Var}(H) = \frac{n \cdot \text{Var}(X)}{n^2} \leq \frac{4}{n}. \end{aligned}$$

2. From the previous part, we know that $\sigma_H \leq 2/\sqrt{n}$. In order to guarantee that it is at most **0.01**, we solve, $2/\sqrt{n} \leq 0.001$ for n to obtain $n \geq 40000$.

3. We apply the Chebyshev inequality to H , with $\mathbf{E}[H]$ and $\text{Var}(H)$ from part (1):

$$\mathbf{P}(|H - h| \geq 0.02) \leq \frac{\sigma_H^2}{0.02^2} \quad \text{or} \quad \mathbf{P}(|H - h| \leq 0.02) \geq 1 - \frac{\sigma_H^2}{0.02^2}.$$

Substituting in our upper bound on σ_H^2 , we obtain

$$1 - \frac{\sigma_H^2}{0.02^2} \geq 1 - \frac{2^2}{n \cdot 0.02^2}.$$

Hence, to guarantee that our estimate is within **0.02** of the true mean h with probability of at least **99%**, it suffices to have,

$$1 - \frac{2^2}{n \cdot 0.02^2} \geq 0.96.$$

Solving this for n , we have that n must satisfy,

$$n \geq 250000.$$

4. Since X is uniform in the interval $[h - 3, h + 3]$, we know that the expected value of X is h and its variance, denoted by σ_H^2 , is 3. Using the standard normal table, and the Central Limit Theorem, we know that for sufficiently large n ,

$$\mathbf{P}\left(\left|\frac{H - h}{\sigma_H/\sqrt{n}}\right| \leq 1.96\right) \approx 0.95.$$

Hence,

$$\mathbf{P}\left(H - \frac{1.96 \cdot \sqrt{3}}{\sqrt{n}} \leq h \leq H + \frac{1.96 \cdot \sqrt{3}}{\sqrt{n}}\right) \approx 0.95.$$

Therefore, the **95%** confidence interval for h is $\left[H - \frac{1.96 \cdot \sqrt{3}}{\sqrt{n}}, H + \frac{1.96 \cdot \sqrt{3}}{\sqrt{n}}\right]$.