

STA260 Summer 2025 - Tutorial 4

July 2025

Overview:

1. Bias
2. MSE

Summary

Bias and MSE

Given a point estimator $\hat{\theta}$ for the parameter θ , the bias is calculated as $B(\hat{\theta}) = E[\hat{\theta}] - \theta$.

We say a point estimator is unbiased if $E[\hat{\theta}] = \theta$.

The mean square error (MSE) of $\hat{\theta}$ is $MSE(\hat{\theta}) = E((\hat{\theta} - \theta)^2) = \text{Var}(\hat{\theta}) + B(\hat{\theta})^2$.

Question 1

Let X_1, X_2, \dots, X_n denote a random sample from a distribution with the probability density function

$$f(x) = \begin{cases} e^{-(x-\mu)} & \text{if } x \geq \mu \\ 0 & \text{otherwise} \end{cases}$$

μ is the smallest val possible

Let $\hat{\mu} = \text{Min}(X_1, X_2, \dots, X_n)$. Calculate the bias of $\hat{\mu}$.

$$B(\theta) = E(\hat{\theta}) - \theta$$

$$B[\hat{\mu}] = E[\hat{\mu}] - \mu$$

$$f_{\hat{\mu}}(x) = n f_x(x) (1 - F_x(x))^{n-1}$$

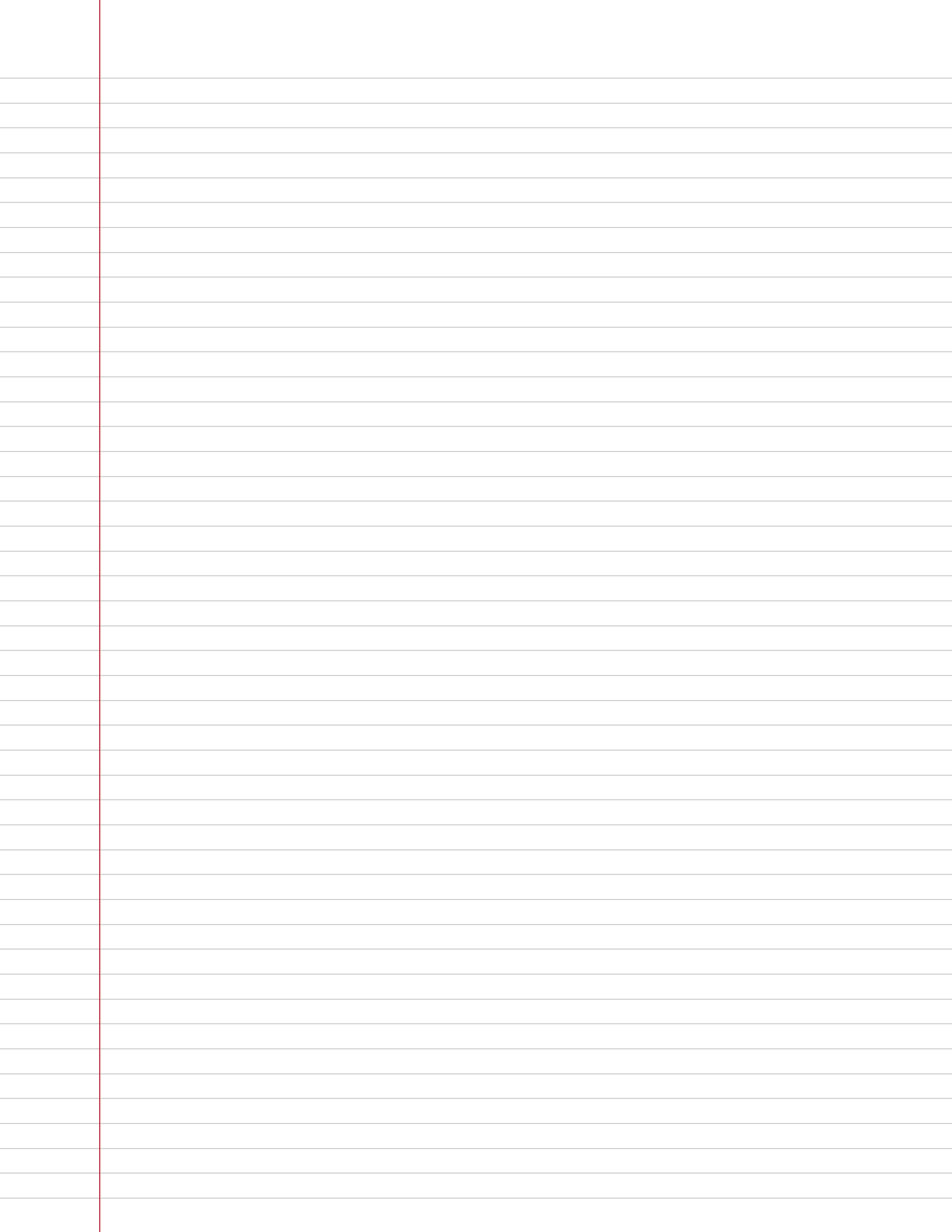
$$= n e^{-(x-\mu)} (1 - e^{-(x-\mu)})^{n-1}$$

$$F_x(x) = \int_{\mu}^x e^{-(t-\mu)} dt = -e^{-(t-\mu)} \Big|_{\mu}^x$$

$$= -e^{-(x-\mu)} + 1$$

$$= n e^{-(x-\mu)} (1 - e^{-(x-\mu)})^{n-1}$$

$$= n e^{-(x-\mu)}$$



Question 2

Let X_1, X_2, \dots, X_n denote a random sample from a distribution with the probability density function

$$f(x) = \begin{cases} e^{-(x-\mu)} & \text{if } x \geq \mu \\ 0 & \text{otherwise} \end{cases}$$

Calculate the mean square error of $\hat{\mu}$.

$$ME(\hat{\mu}) = \text{Var}(\hat{\mu}) + B(\hat{\mu})$$

Question 3

Let Y_1, Y_2, \dots, Y_n denote a random sample from a $\text{Uniform}(\theta, \theta + 1)$ distribution. Consider the following estimators:

$$\hat{\theta}_1 = \bar{Y} - \frac{1}{2} \quad \hat{\theta}_2 = Y_{(n)} - \frac{n}{n+1}$$

Which estimators are unbiased?