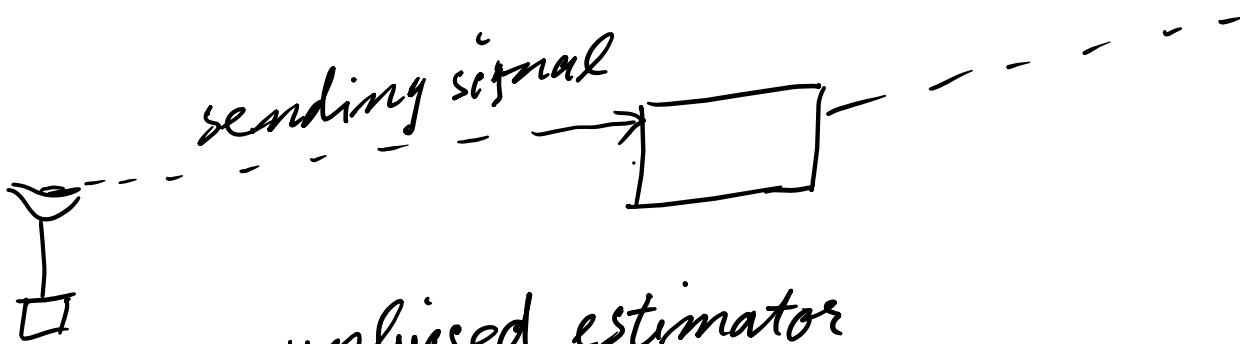


Lecture 11



unbiased estimator

$$M\bar{X} = E(\bar{X}) = M$$

$$\hat{s}^2 = \sigma_{\bar{X}}^2 = \frac{(X_1 - \bar{X})^2 + (X_2 - \bar{X})^2 + \dots + (X_n - \bar{X})^2}{n}$$

\rightarrow biased estimator \rightarrow population mean

$$= \frac{n-1}{n} \sigma^2$$

\rightarrow population variance

$$E(\hat{s}^2) = E\left(\frac{n-1}{n} \sigma^2\right) = \frac{n-1}{n} \cdot \sigma^2$$

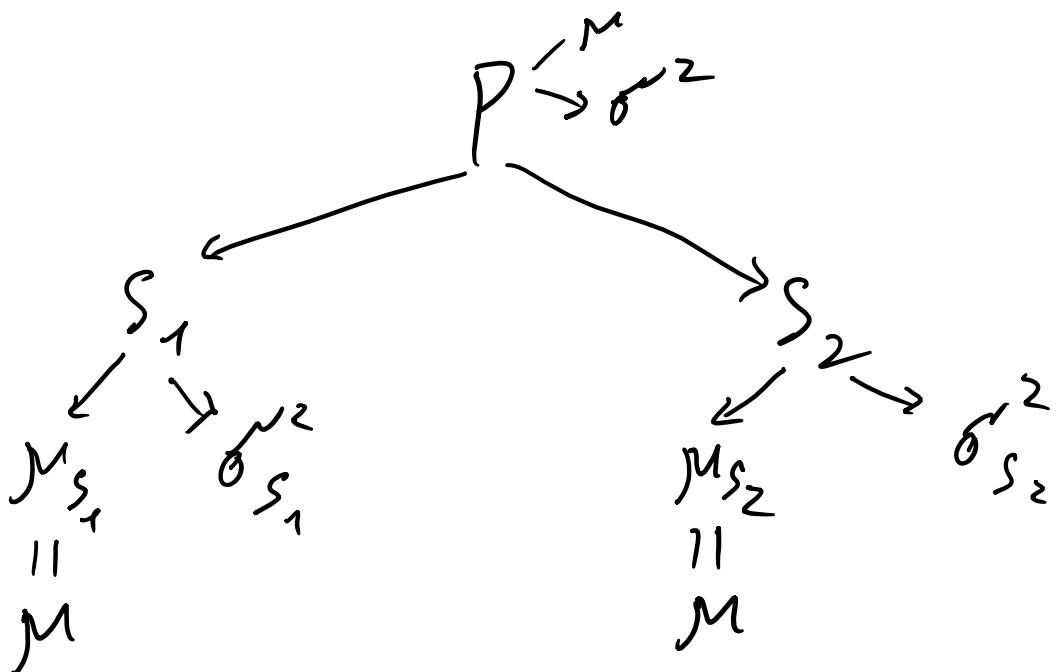
\neq

$$E(\hat{s}^2) = \sigma^2$$

$$\hat{s}^2 = \frac{n}{n-1} s^2 = \frac{n}{n-1} \cdot \frac{n-1}{n} \cdot \sigma^2$$

$$E(\hat{s}^2) = E(s^2) = \sigma^2$$

\rightarrow unbiased estimator



$$\text{If } \sigma_{S_1}^2 < \sigma_{S_2}^2$$

$\rightarrow S_1$ is a more efficient estimator

* Unbiased estimator of standard deviation:

- Normal distribution:

$$E(S) = \sqrt{\frac{2}{n-1}} \cdot \frac{\Gamma\left(\frac{n}{2}\right)}{\Gamma\left(\frac{n-1}{2}\right)} \cdot \sigma$$

$$\Rightarrow \hat{S}! \rightarrow E(\hat{S}) = \sigma$$

- Other distribution:

$$\hat{S} = \sqrt{\frac{1}{n-1,5 - \frac{1}{4} Y_2} \sum_{i=1}^n (x_i - \bar{x})^2}$$

population excess kurtosis

In-class exercise:

Unbiased estimator of mean: 67.45 inches

11

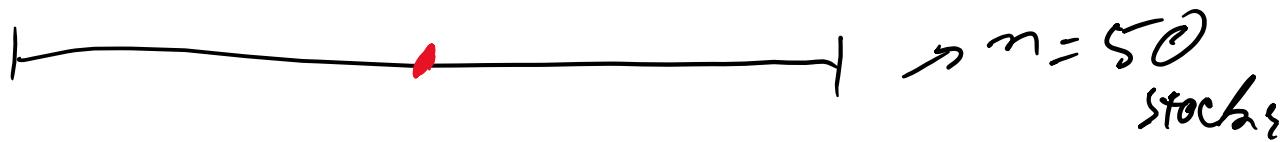
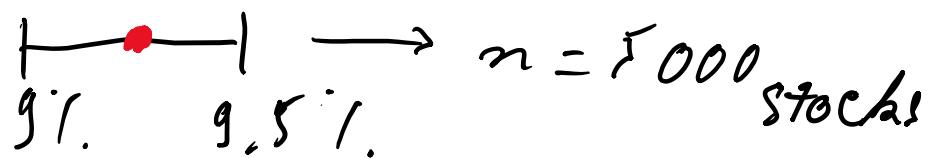
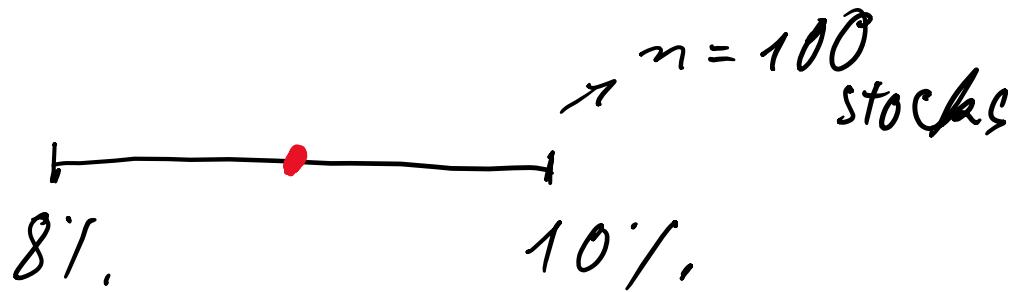
$$\text{Variance} : \frac{100}{99} (8.5275)$$

$$= \hat{s}^2 = 8.61$$

$$\hat{s} = \sqrt{8.61}$$

$$= 2.93$$

Biased estimator
of standard deviation

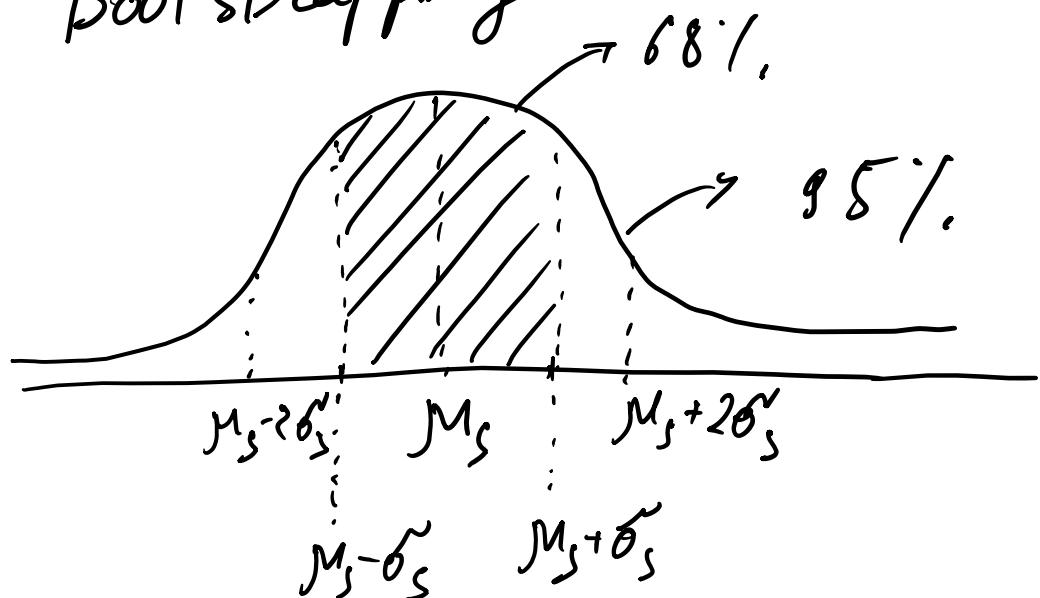


Methods for calculating confidence interval:

- Informal

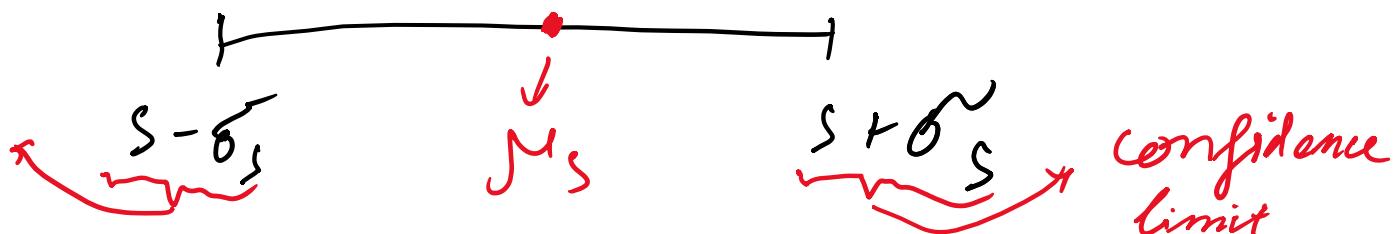
- Normal-based

- Bootstrapping



$[s - \tilde{s}, s + \tilde{s}]$: confidence interval

You are 68%, M_s is lying in this range



$[s - 2\tilde{s}, s + 2\tilde{s}]$

confidence level

95%

confident that M_s is lying in this range

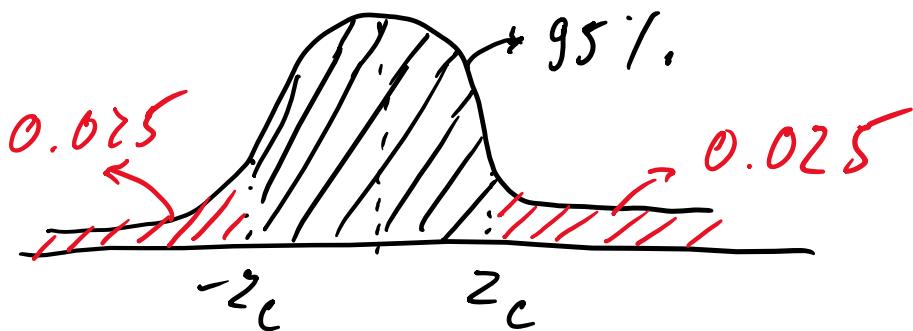


$s - 2\tilde{s}$

M_s

$s + 2\tilde{s}$

Determine the critical value z_c of 95% confidence level



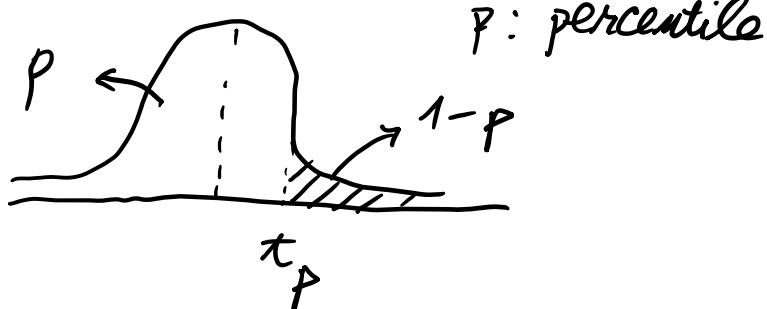
$$0.95/z = 0.475 \rightarrow z_c = 1.96$$

$$99\% \rightarrow z_c = 2.58$$

$$\bar{M}_{\text{Sample}} = 7.5\% \quad 7.4 \quad 7.6$$

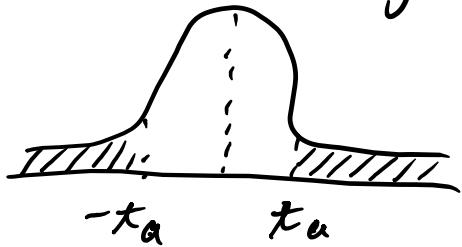
$$\sigma = 1\% \quad \rightarrow \mu = 7.5 \pm \frac{1}{\sqrt{100}} \\ n = 100$$

Student t -distribution



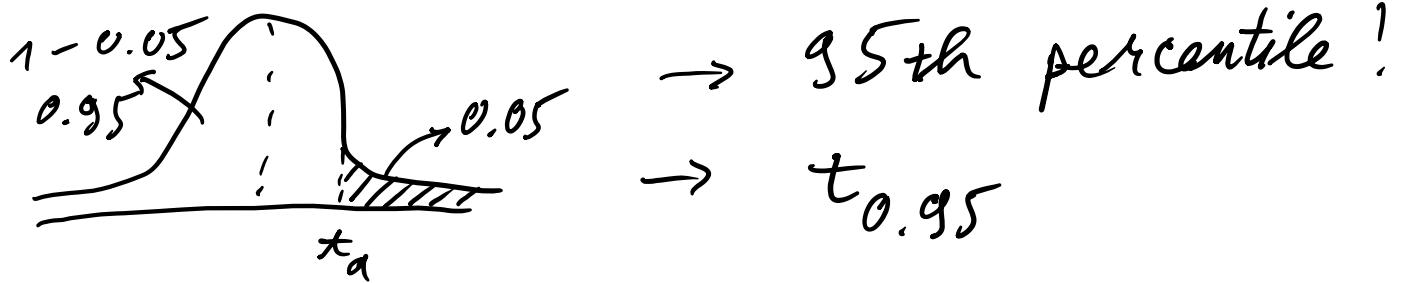
p : percentile

Example: degree of freedom $v = 9$



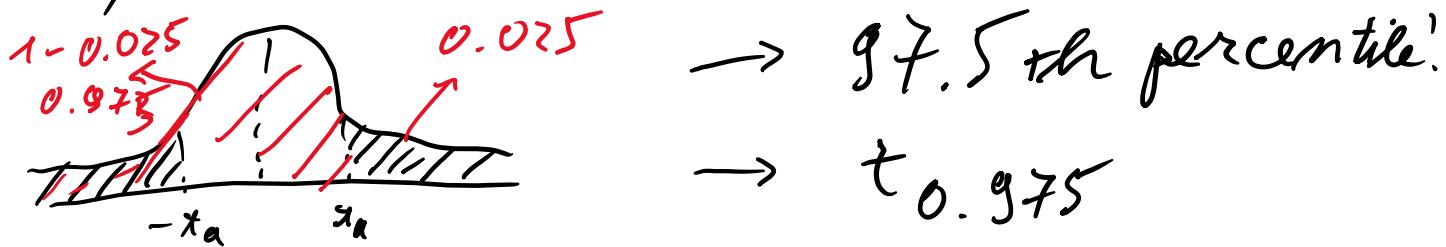
Determine value of t_a if

a) Shaded area on the right = 0.05

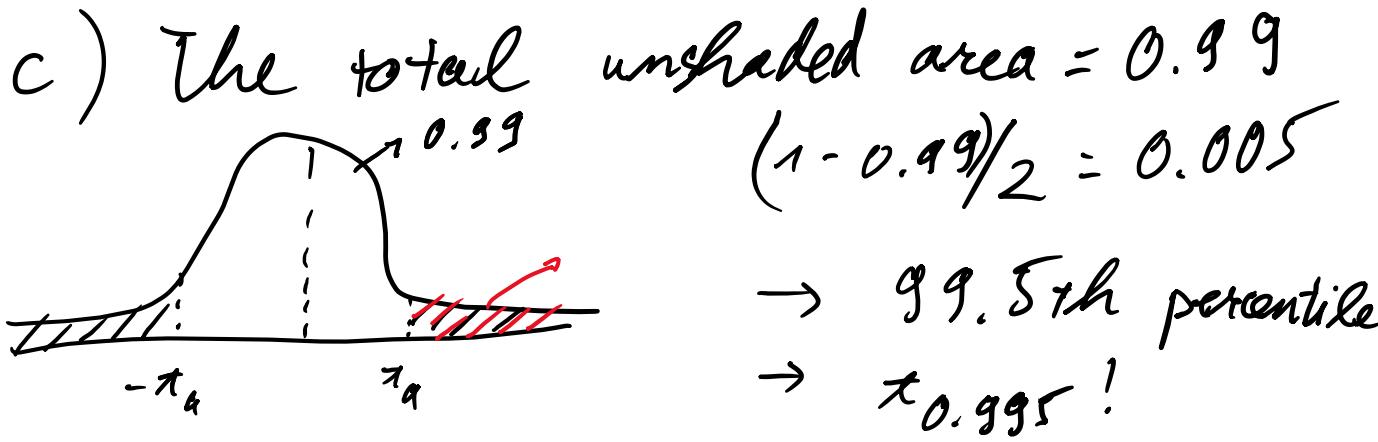


$$t_a = 1.83$$

b) The total shaded area = 0.05



$$t_a = 2.26$$



$$t_\alpha = 3.25$$

In-class exercise:

99% confidence level

$$\rightarrow z_c = 2.58$$

$$\bar{X} = 67.45 \quad \sigma = 2.93 \quad n = 100$$

$$67.45 \pm 2.58 \frac{2.93}{\sqrt{100}} \rightarrow 66.69 < \mu < 68.21$$

$$\bar{X} = 67.45 \quad \sigma = 2.93 \quad n = 20$$

99% confidence level

$$\begin{aligned} v &= n - 1 \\ &= 19 \end{aligned}$$

$$67.45 \pm t_{0.995} \cdot \frac{2.93}{\sqrt{20}}$$

$$\ln(a^x) = x \cdot \ln(a)$$

$$\ln(x \cdot y) = \ln(x) + \ln(y)$$

$$\begin{aligned}\sum_{k=1}^n (x_k - \mu) &= (x_1 - \mu) + (x_2 - \mu) + \dots + (x_n - \mu) \\ &= (x_1 + x_2 + \dots + x_n) - (n\mu) \\ &= \left(\sum_{k=1}^n x_k \right) - n\mu\end{aligned}$$

$$f(x_k, \theta_1, \theta_2) = \frac{1}{\sqrt{2\pi\theta_2}} e^{-\frac{(x_k - \theta_1)^2}{2\theta_2}}$$

$$L = \frac{1}{(2\pi)^{n/2} \theta_2^{n/2}} \cdot \exp\left(-\frac{1}{2\theta_2} \sum_{i=1}^n (x_i - \theta_1)^2\right)$$

$$\ln L = !$$

$$\frac{\partial(\ln L)}{\partial \theta_1} = 0$$

$$\frac{\partial(\ln L)}{\partial \theta_2} = 0$$