

## Motivation

In case that  $\sigma^2$  is unknown we can replace it with  $S_n^2$  for large  $n$

We know that for large  $n$ :  $\frac{\bar{X}_n - \mu}{S_n / \sqrt{n}} \approx N(0, 1)$

In practice, already for  $n \geq 30$  we can compute Z-score confidence intervals

- But, what do we do for small  $n < 30$ ?

If  $X_1 \sim N(\mu, \sigma^2)$  then  $\frac{X_n - \mu}{S_n / \sqrt{n}} \sim t(n - 1)$

t distribution with  $n - 1$  degrees of freedom

For normal population and small  $n$  we can use  $t_{\alpha/2}$  instead of  $z_{\alpha/2}$

We obtain then  $1 - \alpha$  T-score confidence interval:  $\bar{X}_n \pm t_{\alpha/2} \frac{S_n}{\sqrt{n}}$

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## Example

Suppose  $n = 4$  radar guns are set up along a stretch of road to catch people driving over the speed limit. Each radar gun is known to have a normal measurement error  $N(0, \sigma^2)$  with  $\sigma^2$  unknown. For a car passing at speed  $\mu$  four readings are (45.71, 47.41, 40.95, 50.65). Compute a random interval that covers the true unknown car speed  $\mu$  with probability of 0.95.

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For our values, we get  $\alpha = 0.05$ ,  $S_n = 4.04$ ,  $t_{\alpha/2} = 3.18$ ,  $\bar{X}_n = 46.18$  and the 95% confidence interval is (39.74, 52.62).

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