## 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:  $\frac{\overline{X}_n - \mu}{S_n / \sqrt{n}} \approx N(0, 1)$ 

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  $t_{lpha/2}$  instead of  $z_{lpha/2}$ 

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

## 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.

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