

## Definitions and Example for Confidence Intervals

Definition: A  $(1-\alpha)*100\%$  confidence interval for a parameter  $\theta$  is a pair of random variables A and B such that  $P(A \leq \theta \leq B) = 1-\alpha$ .

- A and B are random variables because they depend on sample data.
- Based on a particular sample, we observe realized values  $a$  and  $b$ . The interval  $[a, b]$  is our observed confidence interval.

### Notes:

1) If we want a 95% C.I., then  $\alpha = 0.05$ :

$$(1-\alpha)*100\% = (1-0.05)*100\% = 0.95*100\% = 95\%$$

If we want a 99% C.I., then  $\alpha = 0.01$ .

2) The quantity  $(1-\alpha)*100\%$  (ex 95%) is referred to as the confidence level of the interval.

Example:  $X_1, \dots, X_n \stackrel{iid}{\sim} \text{Normal}(\mu, \sigma^2)$   
 (both  $\mu$  and  $\sigma^2$  unknown)

From day 2 of class:

$$\frac{\bar{X} - \mu}{S/\sqrt{n}} \sim t_{n-1}$$

where  $\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i$  is the sample mean (as a random variable)

and  $S = \sqrt{\frac{\sum_{i=1}^n (X_i - \bar{X})^2}{n-1}}$  is sample standard deviation (as a random variable)

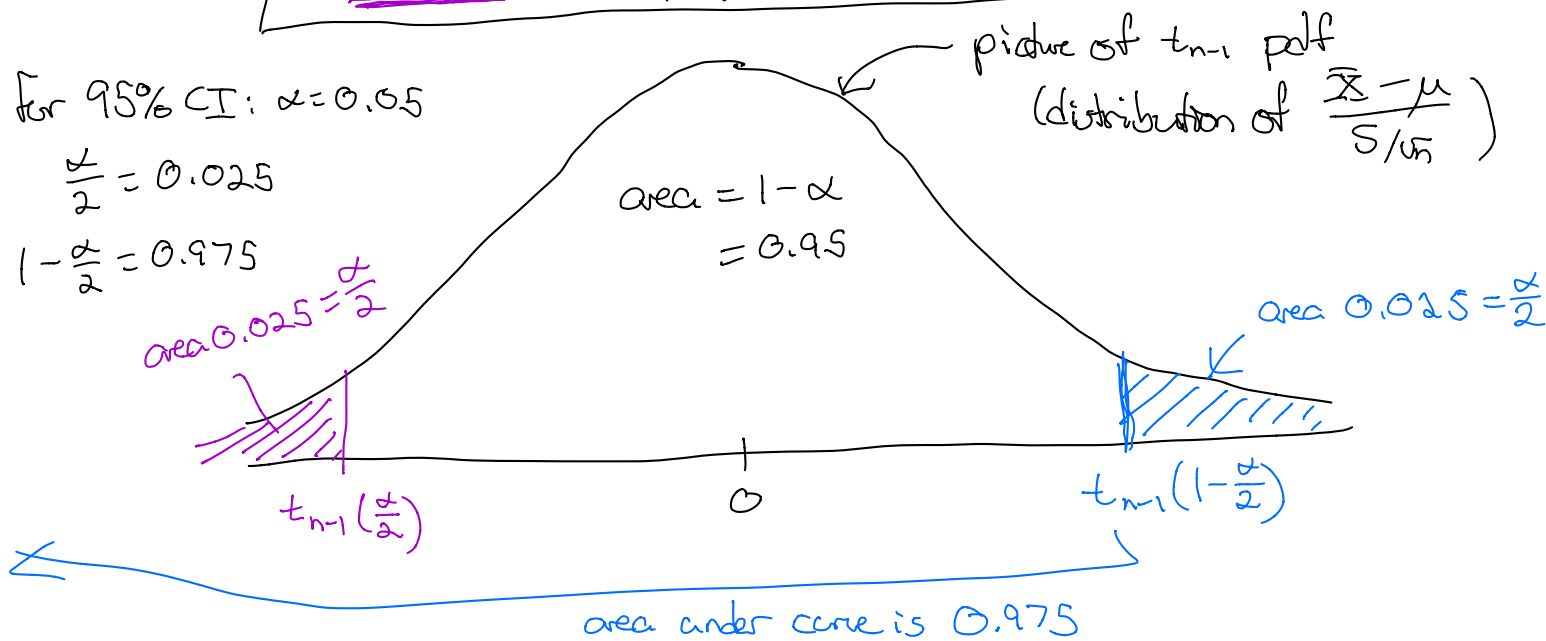
Denote the quantile  $q$  of the  $t_{n-1}$  distribution by  $t_{n-1}(q)$

$$\text{Then } P\left(t_{n-1}\left(\frac{\alpha}{2}\right) \leq \frac{\bar{X} - \mu}{S/\sqrt{n}} \leq t_{n-1}\left(1 - \frac{\alpha}{2}\right)\right) = 1 - \alpha$$

for 95% CI:  $\alpha = 0.05$

$$\frac{\alpha}{2} = 0.025$$

$$1 - \frac{\alpha}{2} = 0.975$$



Our Goal: A pair of random variables  $A, B$  such that  
 $P(A \leq \mu \leq B) = 1 - \alpha$

$$1 - \alpha = P\left(t_{n-1}\left(\frac{\alpha}{2}\right) \leq \frac{\bar{X} - \mu}{S/\sqrt{n}} \leq t_{n-1}\left(1 - \frac{\alpha}{2}\right)\right)$$

random variables  
everything else is a number.

$$= P\left(t_{n-1}\left(\frac{\alpha}{2}\right) \cdot \frac{S}{\sqrt{n}} \leq \bar{X} - \mu \leq t_{n-1}\left(1 - \frac{\alpha}{2}\right) \cdot \frac{S}{\sqrt{n}}\right)$$

$$= P\left(-\bar{X} + t_{n-1}\left(\frac{\alpha}{2}\right) \cdot \frac{S}{\sqrt{n}} \leq -\mu \leq -\bar{X} + t_{n-1}\left(1 - \frac{\alpha}{2}\right) \cdot \frac{S}{\sqrt{n}}\right)$$

$$= P\left(\bar{X} - t_{n-1}\left(\frac{\alpha}{2}\right) \frac{S}{\sqrt{n}} \geq \mu \geq \bar{X} - t_{n-1}\left(1 - \frac{\alpha}{2}\right) \frac{S}{\sqrt{n}}\right)$$

$$= P\left(\underbrace{\bar{X} - t_{n-1}\left(1 - \frac{\alpha}{2}\right) \frac{S}{\sqrt{n}}}_A \leq \mu \leq \underbrace{\bar{X} - t_{n-1}\left(\frac{\alpha}{2}\right) \frac{S}{\sqrt{n}}}_B\right)$$

$$\left. \begin{array}{l} -5 \leq -3 \\ 5 \geq 3 \\ \downarrow \\ 3 \leq 5 \end{array} \right\}$$

Our  $(1 - \alpha) \times 100\%$  confidence interval for  $\mu$  is

$$\left[ \bar{X} - t_{n-1}\left(1 - \frac{\alpha}{2}\right) \frac{S}{\sqrt{n}}, \bar{X} - t_{n-1}\left(\frac{\alpha}{2}\right) \frac{S}{\sqrt{n}} \right]$$

Based on a data set with observed sample mean  $\bar{x}$  and sample standard deviation  $s$ , our observed confidence interval is

$$\left[ \bar{x} - t_{n-1}\left(1 - \frac{\alpha}{2}\right) \frac{s}{\sqrt{n}}, \bar{x} - t_{n-1}\left(\frac{\alpha}{2}\right) \frac{s}{\sqrt{n}} \right]$$

no random variables! all just numbers,  
can't make probability statements.

## Example: Body Temperatures

- It's generally believed that the average body temperature is 98.6 degrees Fahrenheit (37 degrees Celsius).
- Let's investigate with measurements of the temperatures of 130 adults.

### Confidence Interval Calculation

$$\left[ \bar{x} - t_{n-1}\left(1-\frac{\alpha}{2}\right)\frac{s}{\sqrt{n}}, \bar{x} + t_{n-1}\left(\frac{\alpha}{2}\right)\frac{s}{\sqrt{n}} \right]$$

Sample mean and standard deviation:

```
x_bar <- mean(bodytemp$temp)
x_bar
```

```
## [1] 98.24923
```

```
s <- sd(bodytemp$temp)
s
```

```
## [1] 0.7331832
```

Verifying our sample size is  $n = 130$

```
n <- nrow(bodytemp)
n
```

```
## [1] 130
```

Finding appropriate quantiles for a 95% confidence interval:

```
t_lower <- qt(0.025, df = n - 1)
t_lower
```

```
## [1] -1.978524
```

```
t_upper <- qt(0.975, df = n - 1)
t_upper
```

```
## [1] 1.978524
```

Calculation of the confidence interval:

```
x_bar - t_upper * s / sqrt(n) → 98.249 - 1.979 * 0.733 / √130
```

```
## [1] 98.122
```

```
x_bar - t_lower * s / sqrt(n) → 98.249 - (-1.979) * 0.733 / √130
```

```
## [1] 98.37646
```

$$98.249 + 1.979 * 0.733 / \sqrt{130}$$

Our observed confidence interval for  $\mu$  is  $[98.122, 98.376]$

Interpretation: We are 95% confident that the population mean body temperature is between 98.122 °F and 98.376 °F.

Non-interpretation (can't say this!!!)

There is probability 0.95 that  $\mu$  is between 98.122 and 98.376. (This is for a Bayesian credible interval!)

# Comparison of Frequentist confidence intervals and Bayesian credible intervals.

In both cases: a range of plausible values for the unknown parameter  $\theta$ .

	Bayesian	Frequentist
The random variable is:	$\theta$ (the parameter)	A, B the endpoints of the confidence interval
Why is the random variable random?	Expresses our state of knowledge about $\theta$ .	Each sample we take gives a different confidence interval.
How to interpret:	The probability that $\theta$ is in the interval is 0.95	<p>Before taking sample:</p> <ul style="list-style-type: none"> <li>The probability that <math>\theta</math> is in the random interval <math>[A, B]</math> is 0.95</li> </ul> <p>After taking sample:</p> <ul style="list-style-type: none"> <li><math>\theta</math> is either in the interval <math>[a, b]</math> or not, and we don't know for sure whether or not it is.</li> <li>For 95% of samples, the interval calculated based on that sample would contain <math>\theta</math>.</li> </ul>