

# MTH 3210 Lab 3 (Takehome)

Can be handed in Tue., Apr. 30

## 1 Part A: Confidence Interval for $\mu$

### 1.1 Change in Blood Pressure Data Set

If a random sample is drawn from a  $N(\mu, \sigma)$  distribution, then a  $(1 - \alpha)100\%$  *confidence interval* for  $\mu$  is

$$\bar{X} \pm t_{\alpha/2} \frac{S}{\sqrt{n}},$$

where  $t_{\alpha/2}$  is the  $(1 - \alpha/2)100\text{th}$  *percentile* of the  $t$  distribution with  $n - 1$  degrees of freedom.

(The confidence interval is also (approximately) valid if the sample is from a *non-normal* distribution *as long as  $n$  is large*.)

A study was designed to see if increased dietary calcium intake reduces blood pressure. Ten men were given a calcium supplement for 12 weeks. Blood pressure was measured before and after the twelve-week period. The changes in blood pressure for the ten subjects are below:

-7, -5, -5, -17, 8, 5, -1, -10, -11, 2

A **negative** value means the blood pressure **decreased**.

1. Use `c()` create a vector that stores the data.
2. Check the assumption that the sample is from a *normal* distribution by looking at a histogram of the data (using `hist()` after storing the data as a vector).
3. The function `qt()` will return percentiles (or *quantiles*) of a  $t$  distribution. It takes arguments:

<code>p</code>	the probability to the left of the desired $t$ distribution percentile.
<code>df</code>	the degrees of freedom of the $t$ distribution.

For example, to obtain  $t_{0.025}$  (the **97.5th percentile**) from the  $t$  distribution with **nine** degrees of freedom, we type:

```
qt(p = 0.975, df = 9)
```

```
## [1] 2.262157
```

Compute the critical values  $t_{0.005}$ ,  $t_{0.025}$ , and  $t_{0.05}$  from the  $t$  distribution with **nine** degrees of freedom.

4. The `t.test()` function takes a data vector `x` and computes a **95% one-sample t confidence interval** (and carries out a *one-sample t test*) for a population mean  $\mu$ . Among its arguments are:

<code>x</code>	a data vector.
<code>alternative</code>	the direction for the alternative hypothesis, one of "two.sided", "less", or "greater".
<code>mu</code>	the null hypothesized value for the unknown population mean, with default value 0.
<code>conf.level</code>	the confidence level for a confidence interval for the unknown population mean, with default value 0.95.

The optional argument, `conf.level`, can be used to change the level of confidence. For example, the **90%** confidence interval, based on data in a vector called `my.data`, would be computed by typing:

```
t.test(x = my.data, conf.level = 0.90)
```

Use `t.test()` to compute **90%**, **95%**, and **99%** confidence intervals for the true (unknown) mean change in blood pressure  $\mu$ .

## 2 Part B: Confidence Interval for a Population Proportion $p$

### 2.1 Political Poll Results

If a random sample is from a population of *successes* and *failures*, then a  $(1 - \alpha)100\%$  **confidence interval** for the population **proportion** of successes  $p$  is

$$\hat{P} \pm z_{\alpha/2} \sqrt{\frac{\hat{P}(1 - \hat{P})}{n}},$$

where  $z_{\alpha/2}$  is the  $(1 - \alpha/2)100\%$  **percentile** the  $N(0, 1)$  distribution.

(A slightly more accurate version of the CI, called the **score CI**, is given in the textbook.)

A November 18, 2015, online CNN article had the headline "Poll: Clinton trails GOP rivals in Colorado."

The headline referred to a Quinnipiac poll which found that "Dr. Ben Carson tops Democratic front-runner Hillary Clinton **52 - 38** percent in a general election matchup."

Referring to its survey methodology, Quinnipiac states "This RDD (Random Digital Dialing) telephone survey was conducted from November 11 - 15, 2015 throughout the state of Colorado. Responses are reported for **1,262** self-identified registered voters."

1. The function `prop.test()` takes arguments

`x` the number of successes in the sample.  
`n` the sample size.  
`conf.level` the confidence level for a confidence interval for the unknown population proportion, with default value 0.95.

It returns the *sample proportion*  $\hat{P}$  along with a *95% confidence interval for  $p$* .

Assume that **656** (52%) of the **1,262** respondents support Carson, and use `prop.test()` to compute a *95% confidence interval* for the true (unknown) population **proportion**  $p$  that supports him.