

DATA 606 Lab5

Ahsanul Choudhury

October 30, 2016

North Carolina births

In 2004, the state of North Carolina released a large data set containing information on births recorded in this state. This data set is useful to researchers studying the relation between habits and practices of expectant mothers and the birth of their children. We will work with a random sample of observations from this data set.

Exploratory analysis

Load the `nc` data set into our workspace.

```
load("more/nc.RData")
```

We have observations on 13 different variables, some categorical and some numerical. The meaning of each variable is as follows.

variable	description
<code>fage</code>	father's age in years.
<code>mage</code>	mother's age in years.
<code>mature</code>	maturity status of mother.
<code>weeks</code>	length of pregnancy in weeks.
<code>premie</code>	whether the birth was classified as premature (premie) or full-term.
<code>visits</code>	number of hospital visits during pregnancy.
<code>marital</code>	whether mother is married or not married at birth.
<code>gained</code>	weight gained by mother during pregnancy in pounds.
<code>weight</code>	weight of the baby at birth in pounds.

variable	description
lowbirthweight	whether baby was classified as low birthweight (low) or not (not low).
gender	gender of the baby, female or male.
habit	status of the mother as a nonsmoker or a smoker.
whitemom	whether mom is white or not white.

1. What are the cases in this data set? How many cases are there in our sample?

As a first step in the analysis, we should consider summaries of the data. This can be done using the `summary` command:

```
summary(nc)
```

```
##          fage          mage          mature          weeks
##  Min.    :14.00  Min.    :13  mature mom :133  Min.    :20.00
##  1st Qu.:25.00  1st Qu.:22  younger mom:867  1st Qu.:37.00
##  Median :30.00  Median :27                    Median :39.00
##  Mean   :30.26  Mean   :27                    Mean   :38.33
##  3rd Qu.:35.00  3rd Qu.:32                    3rd Qu.:40.00
##  Max.   :55.00  Max.   :50                    Max.   :45.00
##  NA's   :171                    NA's    :2
##          premie          visits          marital          gained
##  full term:846  Min.    : 0.0  married   :386  Min.    : 0.00
##  premie    :152  1st Qu.:10.0  not married:613  1st Qu.:20.00
##  NA's      : 2  Median :12.0  NA's      : 1  Median :30.00
##                    Mean   :12.1                    Mean   :30.33
##                    3rd Qu.:15.0                    3rd Qu.:38.00
##                    Max.   :30.0                    Max.   :85.00
##                    NA's   :9                      NA's   :27
##          weight  lowbirthweight  gender          habit
##  Min.    : 1.000  low      :111  female:503  nonsmoker:873
##  1st Qu.: 6.380  not low:889  male   :497  smoker   :126
##  Median : 7.310                    NA's    : 1
##  Mean   : 7.101
##  3rd Qu.: 8.060
##  Max.   :11.750
##
##          whitemom
##  not white:284
##  white    :714
##  NA's     : 2
```

```
##
##
##
##
```

As you review the variable summaries, consider which variables are categorical and which are numerical. For numerical variables, are there outliers? If you aren't sure or want to take a closer look at the data, make a graph.

Consider the possible relationship between a mother's smoking habit and the weight of her baby. Plotting the data is a useful first step because it helps us quickly visualize trends, identify strong associations, and develop research questions.

2. Make a side-by-side boxplot of `habit` and `weight`. What does the plot highlight about the relationship between these two variables?

The box plots show how the medians of the two distributions compare, but we can also compare the means of the distributions using the following function to split the `weight` variable into the `habit` groups, then take the mean of each using the `mean` function.

```
by(nc$weight, nc$habit, mean)

## nc$habit: nonsmoker
## [1] 7.144273
## -----
## nc$habit: smoker
## [1] 6.82873
```

There is an observed difference, but is this difference statistically significant? In order to answer this question we will conduct a hypothesis test .

Inference

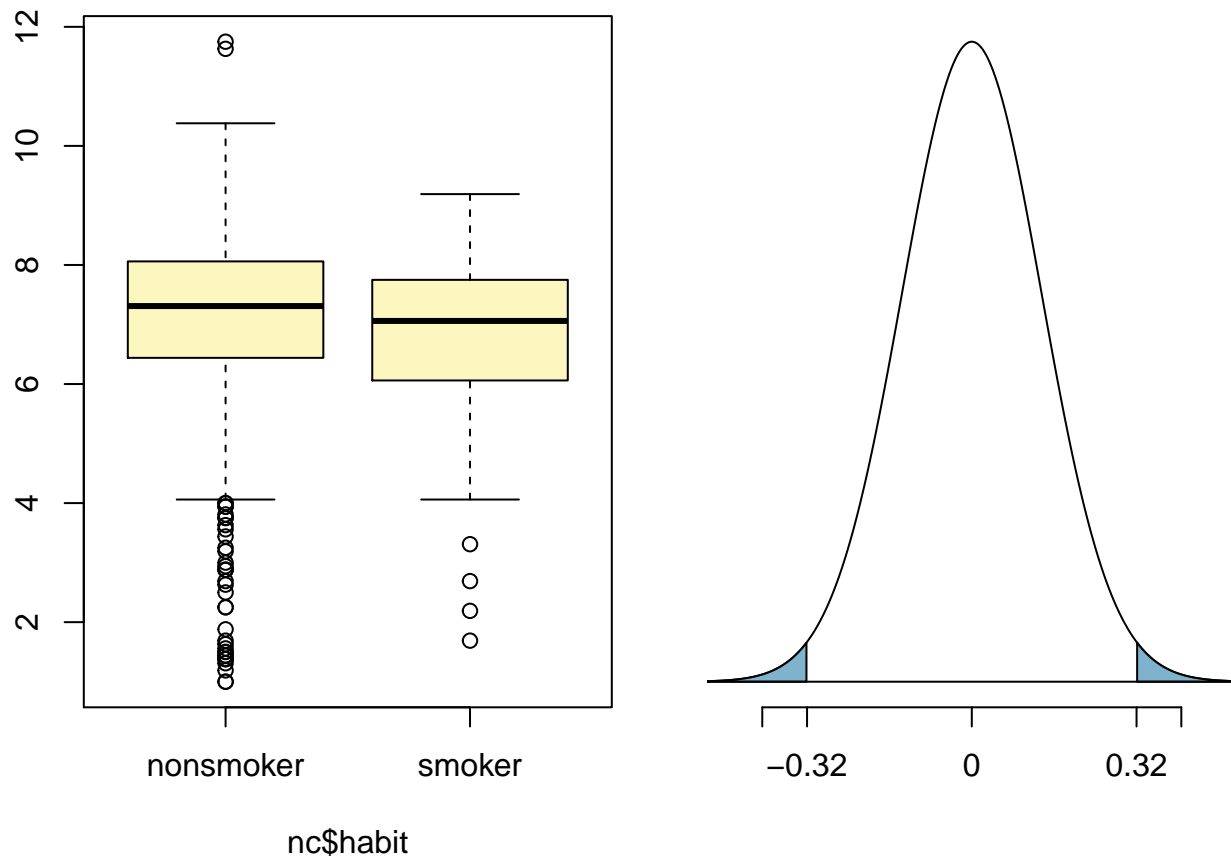
3. Check if the conditions necessary for inference are satisfied. Note that you will need to obtain sample sizes to check the conditions. You can compute the group size using the same `by` command above but replacing `mean` with `length`.
4. Write the hypotheses for testing if the average weights of babies born to smoking and non-smoking mothers are different.

Next, we introduce a new function, `inference`, that we will use for conducting hypothesis tests and constructing confidence intervals.

```
inference(y = nc$weight, x = nc$habit, est = "mean", type = "ht", null = 0,
          alternative = "twosided", method = "theoretical")

## Response variable: numerical, Explanatory variable: categorical
## Difference between two means
## Summary statistics:
## n_nonsmoker = 873, mean_nonsmoker = 7.1443, sd_nonsmoker = 1.5187
## n_smoker = 126, mean_smoker = 6.8287, sd_smoker = 1.3862
```

```
## Observed difference between means (nonsmoker-smoker) = 0.3155
##
## H0: mu_nonsmoker - mu_smoker = 0
## HA: mu_nonsmoker - mu_smoker != 0
## Standard error = 0.134
## Test statistic: Z = 2.359
## p-value = 0.0184
```



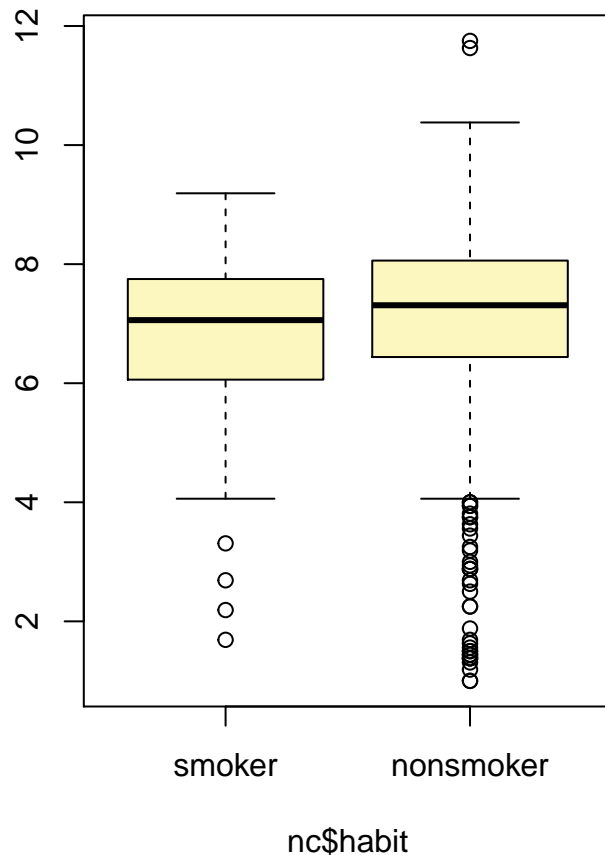
Let's pause for a moment to go through the arguments of this custom function. The first argument is `y`, which is the response variable that we are interested in: `nc$weight`. The second argument is the explanatory variable, `x`, which is the variable that splits the data into two groups, smokers and non-smokers: `nc$habit`. The third argument, `est`, is the parameter we're interested in: `"mean"` (other options are `"median"`, or `"proportion"`.) Next we decide on the `type` of inference we want: a hypothesis test (`"ht"`) or a confidence interval (`"ci"`). When performing a hypothesis test, we also need to supply the null value, which in this case is 0, since the null hypothesis sets the two population means equal to each other. The `alternative` hypothesis can be `"less"`, `"greater"`, or `"twosided"`. Lastly, the `method` of inference can be `"theoretical"` or `"simulation"` based.

5. Change the `type` argument to `"ci"` to construct and record a confidence interval for the difference between the weights of babies born to smoking and non-smoking mothers.

By default the function reports an interval for $(\mu_{nonsmoker} - \mu_{smoker})$. We can easily change this order by using the `order` argument:

```
inference(y = nc$weight, x = nc$habit, est = "mean", type = "ci", null = 0,
          alternative = "twosided", method = "theoretical", order = c("smoker", "nonsmoker"))
```

```
## Response variable: numerical, Explanatory variable: categorical
## Difference between two means
## Summary statistics:
## n_smoker = 126, mean_smoker = 6.8287, sd_smoker = 1.3862
## n_nonsmoker = 873, mean_nonsmoker = 7.1443, sd_nonsmoker = 1.5187
```



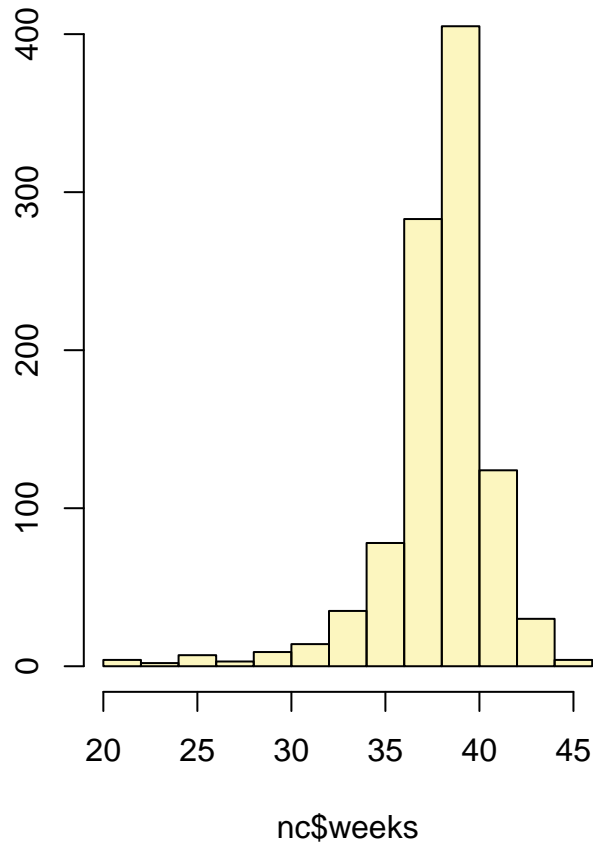
```
## Observed difference between means (smoker-nonsmoker) = -0.3155
##
## Standard error = 0.1338
## 95 % Confidence interval = ( -0.5777 , -0.0534 )
```

On your own

- Calculate a 95% confidence interval for the average length of pregnancies (**weeks**) and interpret it in context. Note that since you're doing inference on a single population parameter, there is no explanatory variable, so you can omit the x variable from the function.

```
inference(y = nc$weeks, est = "mean", type = "ci", null = 0, alternative = "twosided",
  method = "theoretical")
```

```
## Single mean
## Summary statistics:
```



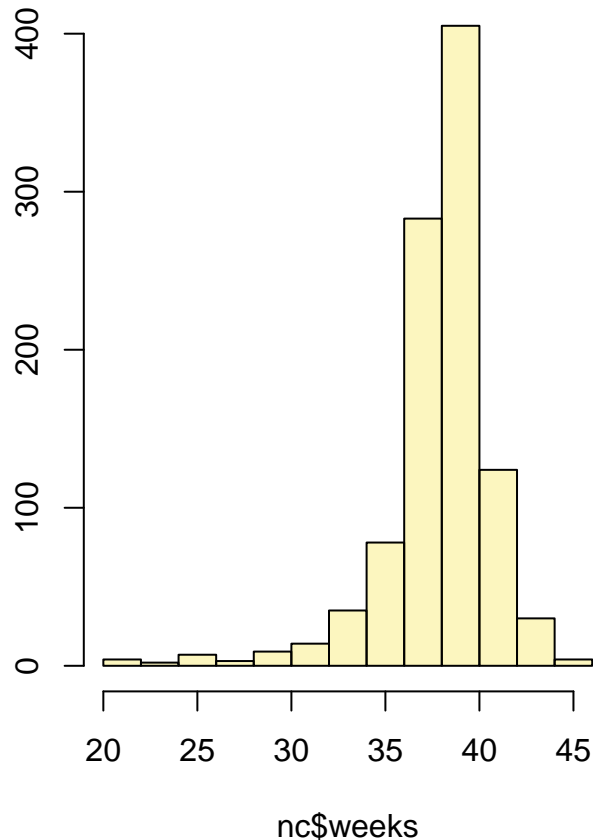
```
## mean = 38.3347 ; sd = 2.9316 ; n = 998
## Standard error = 0.0928
## 95 % Confidence interval = ( 38.1528 , 38.5165 )
```

The confidence interval is relatively small and is huddled around 38, the distribution is left skewed.

- Calculate a new confidence interval for the same parameter at the 90% confidence level. You can change the confidence level by adding a new argument to the function: `conflvel = 0.90`.

```
inference(y = nc$weeks, est = "mean", type = "ci", null = 0, alternative = "twosided",
  method = "theoretical", conflvel = 0.9)
```

```
## Single mean
## Summary statistics:
```



```
## mean = 38.3347 ; sd = 2.9316 ; n = 998
## Standard error = 0.0928
## 90 % Confidence interval = ( 38.182 , 38.4873 )
```

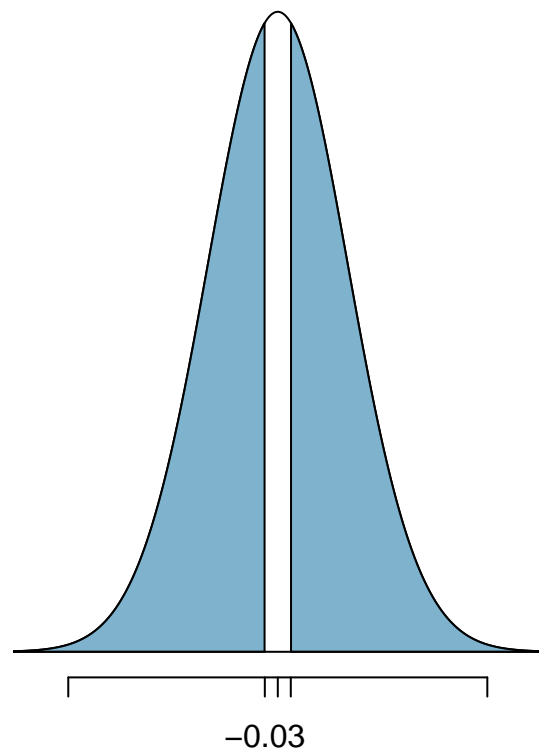
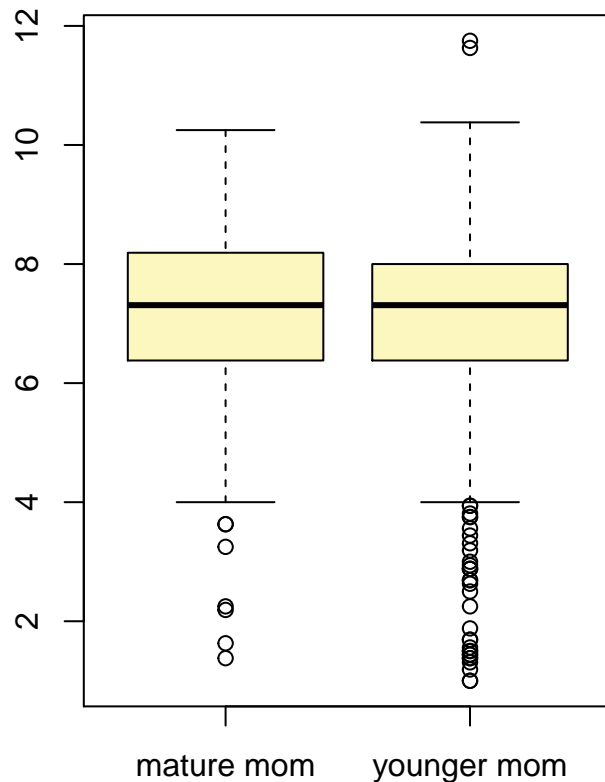
At 90% confidence interval the range gets smaller as 90% of the mean falls within the interval instead of 95%.

- Conduct a hypothesis test evaluating whether the average weight gained by younger mothers is different than the average weight gained by mature mothers.

```
inference(y = nc$weight, x = nc$mature, est = "mean", type = "ht", null = 0,
          alternative = "twosided", method = "theoretical")
```

```
## Response variable: numerical, Explanatory variable: categorical
## Difference between two means
## Summary statistics:
## n_mature mom = 133, mean_mature mom = 7.1256, sd_mature mom = 1.6591
## n_younger mom = 867, mean_younger mom = 7.0972, sd_younger mom = 1.4855

## Observed difference between means (mature mom-younger mom) = 0.0283
##
## H0: mu_mature mom - mu_younger mom = 0
## HA: mu_mature mom - mu_younger mom != 0
## Standard error = 0.152
## Test statistic: Z = 0.186
## p-value = 0.8526
```



nc\$mature

The p-value is 0.8526 which is higher than 0.05, therefore we cannot reject the null hypothesis, the difference between the means is also not statistically significant which leads us to conclusion that there are no difference in average weight gained by younger mothers and mature mothers.

- Now, a non-inference task: Determine the age cutoff for younger and mature mothers. Use a method of your choice, and explain how your method works.

To determine the age cutoff for younger and mature mothers I am going to find the maximum age of mother (mage) where the maturity status (mature) is equal to “younger mom”. Alternatively I can also look for the minimum age of mother (mage) where the maturity status is “mature mom”.

```
max(nc$mage[nc$mature == "younger mom"])
```

```
## [1] 34
```

```
min(nc$mage[nc$mature == "mature mom"])
```

```
## [1] 35
```

Based on my method 34 years and under considered “younger mom” and 35 years and higher considered “mature mom” in our data set.

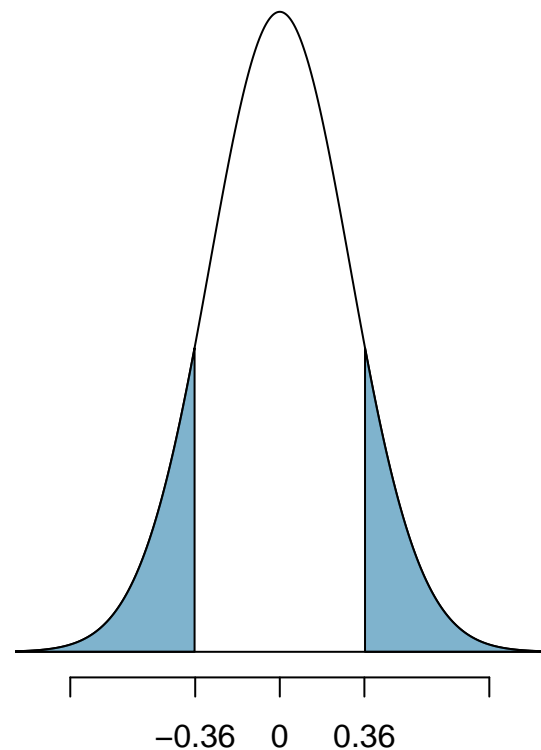
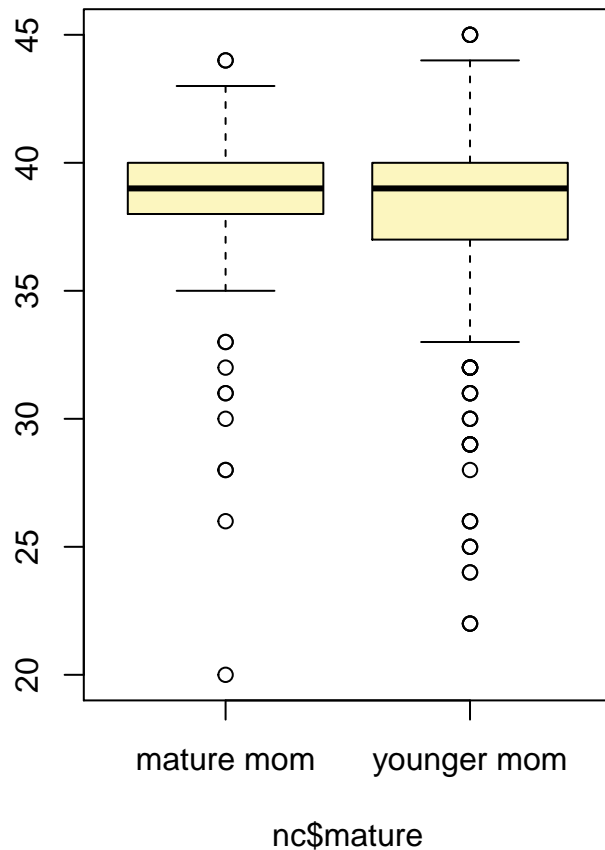
- Pick a pair of numerical and categorical variables and come up with a research question evaluating the relationship between these variables. Formulate the question in a way that it can be answered using a hypothesis test and/or a confidence interval. Answer your question using the **inference** function, report the statistical results, and also provide an explanation in plain language.

Research Question : Is there a relationship between mother's age maturity and pregnancy length in weeks?
 In my research weeks will be numerical and mature will be categorical variable. H0: There is no impact of mother's age maturity on duration of pregnancy. HA: There is a impact of mother's age maturity on duration of pregnancy.

```
inference(y = nc$weeks, x = nc$mature, est = "mean", type = "ht", null = 0,
          alternative = "twosided", method = "theoretical")
```

```
## Response variable: numerical, Explanatory variable: categorical
## Difference between two means
## Summary statistics:
## n_mature mom = 132, mean_mature mom = 38.0227, sd_mature mom = 3.2184
## n_younger mom = 866, mean_younger mom = 38.3822, sd_younger mom = 2.8844

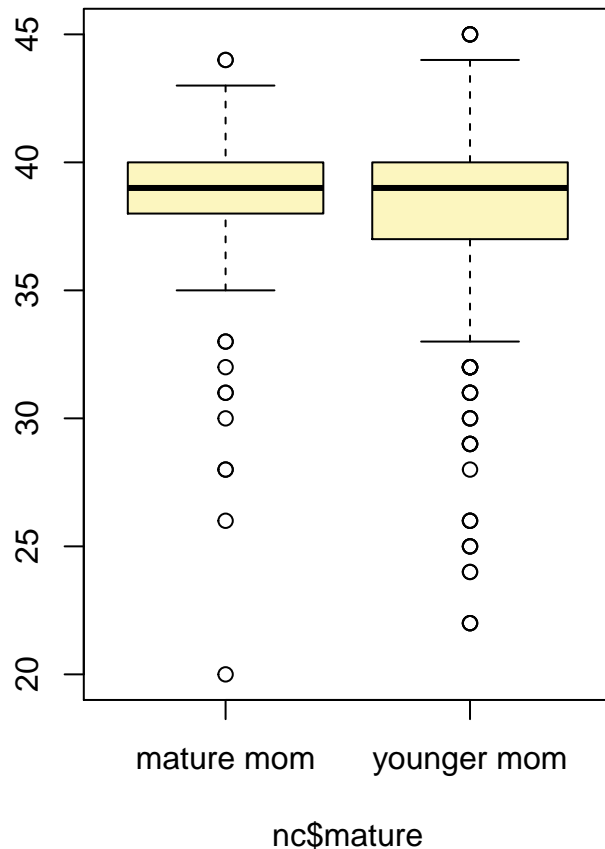
## Observed difference between means (mature mom-younger mom) = -0.3595
##
## H0: mu_mature mom - mu_younger mom = 0
## HA: mu_mature mom - mu_younger mom != 0
## Standard error = 0.297
## Test statistic: Z = -1.211
## p-value = 0.2258
```



The p-value is at 0.2258 which is higher than 0.05, therefore we cannot reject null hypothesis, the statistically insignificant difference in the mean also confirms this conclusion.

```
inference(y = nc$weeks, x = nc$mature, est = "mean", type = "ci", null = 0,
          alternative = "twosided", method = "theoretical")
```

```
## Response variable: numerical, Explanatory variable: categorical
## Difference between two means
## Summary statistics:
## n_mature mom = 132, mean_mature mom = 38.0227, sd_mature mom = 3.2184
## n_younger mom = 866, mean_younger mom = 38.3822, sd_younger mom = 2.8844
```



```
## Observed difference between means (mature mom-younger mom) = -0.3595
##
## Standard error = 0.2968
## 95 % Confidence interval = ( -0.9412 , 0.2222 )
```

The 90% confidence interval contains 0 therefore we cannot reject null hypothesis. Both hypothesis test and confidence interval test results confirms there is no impact of mother's age maturity level on pregnancy duration.

This is a product of OpenIntro that is released under a Creative Commons Attribution-ShareAlike 3.0 Unported. This lab was adapted for OpenIntro by Mine Çetinkaya-Rundel from a lab written by the faculty and TAs of UCLA Statistics.