# Statistical Inference Project Part-2

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### Introduction

This report aims to analyze the ToothGrowth data in the R datasets package. Per the course project instructions, the following items should occur:

- 1. Load the ToothGrowth data and perform some basic exploratory data analyses
- 2. Provide a basic summary of the data.
- 3. Use confidence intervals and/or hypothesis tests to compare tooth growth by supp and dose. (Only use the techniques from class, even if there's other approaches worth considering)
- 4. State your conclusions and the assumptions needed for your conclusions.

# Analysis

Load Libraries and ToothGrowth Data and investigate its structure.

```
library(datasets)
library(dplyr)
## Warning: package 'dplyr' was built under R version 4.0.2
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
##
       filter, lag
## The following objects are masked from 'package:base':
##
##
       intersect, setdiff, setequal, union
library(ggplot2)
## Warning: package 'ggplot2' was built under R version 4.0.2
# Load ToothGrowth Data
data("ToothGrowth")
str(ToothGrowth)
```

```
## 'data.frame': 60 obs. of 3 variables:
## $ len : num  4.2 11.5 7.3 5.8 6.4 10 11.2 11.2 5.2 7 ...
## $ supp: Factor w/ 2 levels "OJ","VC": 2 2 2 2 2 2 2 2 2 2 2 ...
## $ dose: num  0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 ...
```

### summary(ToothGrowth)

##	len	supp	dose
##	Min. : 4.20	OJ:30	Min. :0.500
##	1st Qu.:13.07	VC:30	1st Qu.:0.500
##	Median :19.25		Median :1.000
##	Mean :18.81		Mean :1.167
##	3rd Qu.:25.27		3rd Qu.:2.000
##	Max. :33.90		Max. :2.000

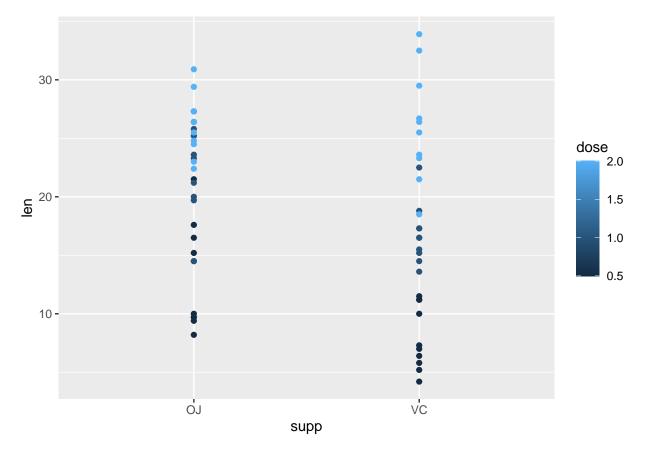
len : Tooth Length is numeric

supp: Supplement is Factor consisting of OJ and VC

dose: Dosage is numeric. As it is just 0.5, 1 and 2, we will have to convert it to factor so that it is easier for analysis

Doing a quick plot to see if there are any patterns.

# qplot(supp,len, data = ToothGrowth, colour = dose)

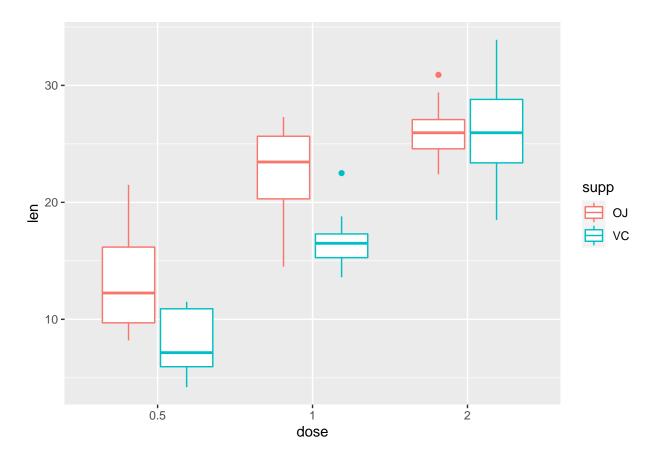


There seems to be some patterns between supp and dose. Let's use boxplot to investigate further.

```
ToothGrowth$dose <- as.factor(ToothGrowth$dose)

# Boxplot for better view

g <- ggplot(ToothGrowth, aes(x=dose, y=len, colour=supp)) + geom_boxplot()
g</pre>
```



We have noticed: 1. There is a trend between dose and len.

2. There is a trend between dose and supp.

### Hypothesis Testing No. 1

1. Hypothesis Testing No.1: Overall impact of dosage on tooth growth. 2. Null Hypothesis - There is no effect of dosage on tooth growth. 3. Alternate Hypothesis - Higer doses produce greater tooth growth.

```
# Arranging the dataset

dose05 <- filter(ToothGrowth, dose == 0.5)
dose1 <- filter(ToothGrowth, dose == 1)
dose2 <- filter(ToothGrowth, dose == 2)</pre>
```

Comparing dosage of 1mg vs 0.5mg

```
# compare between dose of 0.5 and 1

t.test(dose1$len, dose05$len, alternative = "greater")
```

As P value is less than 0.001, it is highly significant and we can reject the null hypothesis

Comparing dosage of 2mg vs 1mg

```
t.test(dose2$len, dose1$len, alternative = "greater")
```

### Conclusion No. 1

As P value is less than 0.001, it is highly significant and we can reject the null hypothesis. So in both casess, we reject the null hypothesis.

We accept the althernative: higher dosages do have a positive influence on toothgrowth.

Hypothesis Testing No. 2

1. Hypothesis Testing No 2: Differences between supplements by dosage. 2. Null Hypothesis - There is no effect between supplement OJ and VC. 3. Alternate Hypothesis - OJ is better on toothgrowth than VC.

We will do three tests and compare them based on dosages.

Dosage: 0.5mg

```
OJdose05 <- filter(dose05, supp == "OJ")
VCdose05 <- filter(dose05, supp == "VC")
t.test(OJdose05$len, VCdose05$len, alternative = "greater")</pre>
```

```
##
## Welch Two Sample t-test
##
## data: OJdose05$len and VCdose05$len
## t = 3.1697, df = 14.969, p-value = 0.003179
## alternative hypothesis: true difference in means is greater than 0
## 95 percent confidence interval:
## 2.34604    Inf
## sample estimates:
## mean of x mean of y
## 13.23    7.98
```

P value is lower than 0.05. We reject the null hypothesis. We accept the alternative: OJ is a better supplement than VC at dosage of  $0.5 \,\mathrm{mg}$ 

#### Dosage: 1mg

```
OJdose1 <- filter(dose1, supp == "OJ")
VCdose1 <- filter(dose1, supp == "VC")
t.test(OJdose1$len, VCdose1$len, alternative = "greater")</pre>
```

P value is lower than 0.05. We reject the null hypothesis. We accept the alternative: OJ is a better supplement than VC at dosage of 1mg

#### Dosage: 2mg

```
OJdose2 <- filter(dose2, supp == "OJ")
VCdose2 <- filter(dose2, supp == "VC")
t.test(OJdose2$len, VCdose2$len, alternative = "greater")</pre>
```

```
## 95 percent confidence interval:
## -3.1335    Inf
## sample estimates:
## mean of x mean of y
## 26.06    26.14
```

P value is higher than 0.05. We cannot reject the null hypothesis: at dosage of 2mg, there is no significant difference between the two supplement

### Conclusion No. 2

For dosages of 0.5mg and 1mg, there are significant differences: OJ is better than VC in promoting tooth growth. At 2mg however, there is no significant differences.