# Investigation of Tooth Growth Data in R

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## Summary

This document describes a statistical inference exercise relating to the effect of supplements on tooth growth in guinea pigs. The tools of statistical inference and the R programming environment will be used to explore the ToothGrowth data set.

Tooth growth in guinea pigs can be said to increase with increasing dose of a supplement (either orange juice or vitamin C) with >95% confidence. However, it is not possible to determine from this data whether orange juice or vitamin C produce a greater effect.

## **Exploratory Data Analysis**

Reading the help file for ToothGrowth provides the following information:

"The response is the length of odontoblasts (teeth) in each of 10 guinea pigs at each of three dose levels of Vitamin C (0.5, 1, and 2 mg) with each of two delivery methods (orange juice or ascorbic acid)."

The following is a quick summary of the data set:

##	len	supp	dose	<pre>factor(ToothGrowth\$dose)</pre>
##	Min. : 4.20	OJ:30	Min. :0.500	0.5:20
##	1st Qu.:13.07	VC:30	1st Qu.:0.500	1 :20
##	Median :19.25		Median :1.000	2 :20
##	Mean :18.81		Mean :1.167	
##	3rd Qu.:25.27		3rd Qu.:2.000	
##	Max. :33.90		Max. :2.000	

There are 60 observations as expected over 10 animals on 2 supplements each given at 3 different doses. There are no NA values.

It's also useful to know from the description that the data is collected from always the same 10 specimens. It is reasonable to infer that any change in tooth growth is a response to the supplement and not a difference between individual animals.

However, a flaw is that the dataset does not identify the individual animals and a search of online references does not give any confirmation or otherwise if the set is ordered. Therefore paired t-tests cannot be relied upon, although paired and unpaired tests could be run for comparison.

## Summarising the analysis so far:

Appendix A shows a violin plot of tooth length comparing the two supplements.

From this graph, there appear to be two factors playing out on the tooth growth.

- 1. There appears to be a clear trend of increasing tooth growth with increasing supplement dose.
- 2. Less clear is whether or not the specific supplement makes a difference (pure vitamin C vs orange juice).

### T-tests and confidence intervals

#### The effect of dose

The effect of dose seems to be intuitive from the graph. Nevertheless, this can be quantified.

For the first analysis, the data are separated into samples according to dose (regardless of supplement) and t-tests are run to compare the 0.5 vs 1 and the 1 vs 2 dose samples. Furthermore, paired and unpaired tests are run for comparison purposes. As previously stated, it is not certain whether the data can be assumed to be ordered and paired.

```
half <- ToothGrowth$len[ToothGrowth$dose == 0.5]
one <- ToothGrowth$len[ToothGrowth$dose == 1]
two <- ToothGrowth$len[ToothGrowth$dose == 2]

tp.5vs1 <- t.test(half, one, paired = TRUE)
tu.5vs1 <- t.test(half, one, paired = FALSE)

tp1vs2 <- t.test(one, two, paired = TRUE)
tu1vs2 <- t.test(one, two, paired = FALSE)</pre>
```

The confidence intervals for dose 0.5 vs dose 1.0, for paired and unpaired tests, respectively are:

```
tp.5vs1$conf.int

## [1] -11.872879 -6.387121
## attr(,"conf.level")
## [1] 0.95

tu.5vs1$conf.int

## [1] -11.983781 -6.276219
## attr(,"conf.level")
## [1] 0.95
```

The confidence intervals for dose 1 vs dose 2, for paired and unpaired tests, respectively are:

```
tp1vs2$conf.int

## [1] -9.258186 -3.471814

## attr(,"conf.level")

## [1] 0.95

tu1vs2$conf.int

## [1] -8.996481 -3.733519

## attr(,"conf.level")

## [1] 0.95
```

In all cases, the confidence interval does not include zero. It doesn't matter if paired or unpaired tests are used. The null hypothesis is rejected. There is >95% confidence that the samples are different. The higher dose gives a higher value of tooth length.

### The effect of supplement

The following demonstrates a t-test on the samples according to supplement (regardless of dose).

```
tpOJvsVC <- t.test(len ~ supp, paired = TRUE, data = ToothGrowth)
tuOJvsVC <- t.test(len ~ supp, paired = FALSE, data = ToothGrowth)</pre>
```

The confidence intervals for OJ vs VC, for paired and unpaired tests, respectively are:

```
tp0JvsVC$conf.int

## [1] 1.408659 5.991341
## attr(,"conf.level")
## [1] 0.95

tu0JvsVC$conf.int

## [1] -0.1710156 7.5710156
## attr(,"conf.level")
## [1] 0.95
```

Here, there is not enough evidence of difference. The unpaired test confidence interval contains zero. There is no evidence that a paired test is valid with this data so the null hypothesis is accepted. The samples may not be different.

## Conclusions

The data clearly show that increasing the dose of supplement increases tooth growth and this is the case whether or not we assume the data sets were paired.

It is not certain that pure vitamin C is better or worse than orange juice because the unpaired test confidence interval contains 0, meaning they might be the same. The null hypothesis is not rejected.

The results from the paired test cannot be relied upon because there is no indication from the data or from external sources that the data a ordered and can be paired.

## Assumptions

t-tests such as shown in the above demonstration rely on a number of assumptions:

- 1. The data should be symmetrically distributed. This is not the case in the guinea pig data set as can be seen from the violin plots. Other tests that the t-test may be more appropriate in these cases, but t-test was selected because it is ideally suited to small samples.
- 2. The pairing relationship between samples must be known. This is not the case in the guinea pig data, and so the accuracy cannot be correctly determined.
- 3. The equality of variance of the samples under comparison must be known (or shall be assumed different). This was beyond the scope of the above demonstration.

# Appendix A

