

Analysis of the Effect of Vitamin C on Tooth Growth in Guinea Pigs

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Synopsys

This report contains exploratory analysis of the tooth growth data and investigates statistical significance of reported effects.

Summary of 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 ...
## $ dose: num 0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 ...
```

Dataset “ToothGrowth” coming with R contains data on the response is the length of odontoblasts (cells responsible for tooth growth) in 60 guinea pigs. Each animal received one of three dose levels of vitamin C (0.5, 1, and 2 mg/day) by one of two delivery methods, orange juice or ascorbic acid (a form of vitamin C and coded as VC).

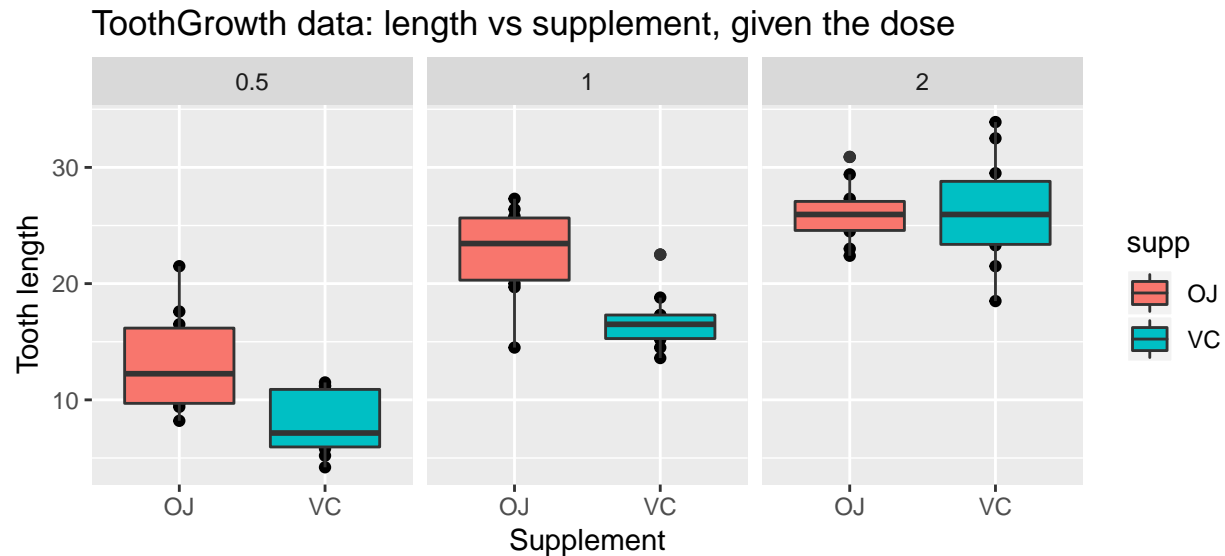
```
summary(ToothGrowth)
```

```
##      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
```

There don't seem to be any missing values and the values of len vary quite a lot. Let's plot the data to see if there are any obvious patterns

```
library(ggplot2)
```

```
qplot(supp, len, data=ToothGrowth, facets=~dose,
      main="ToothGrowth data: length vs supplement, given the dose",
      xlab="Supplement", ylab="Tooth length") +
  geom_boxplot(aes(fill = supp))
```



For the smaller dosages orange juice seems to have more effect on tooth growth than vitamin C, but with the dosage increased to 2 mg the associated tooth growth is very similar.

Assumptions

Since we don't know the underlying population characteristics we will have to make several assumptions:

- Population data is normally distributed
- The variables must be independent and identically distributed (i.i.d.).
- Variances of tooth growth are different when using different supplement and dosage

Hypothesis

Let the H_0 be that the average tooth growth in a given dosage for both supplements are equal, and the alternative H_1 be that orange juice is more effective in smaller dosages (0.5, 1).

For the dosage of 2 mg we'll perform 2-sided test with alternative hypothesis $H_1: \mu_0 <> \mu_1$

```
table(ToothGrowth$supp, ToothGrowth$dose)
```

```
##
##      0.5  1  2
##  OJ   10 10 10
##  VC   10 10 10
```

Each experiment contains exactly 10 observations, so t-test will be appropriate to use.

Let's also assume an acceptable alpha level at 0.05.

```
alpha = 0.05
```

And run the t-tests for different dosages. In an essence we want to see if the difference in means of OJ experiments - VC experiments is greater than zero with 95% confidence.

```
t1 <- split(ToothGrowth, ToothGrowth$dose)
```

```
testres = NULL
```

```
for (i in c(1:2)) {
```

```

      testres[[i]] <- with(tl[[i]],
        t.test(len[supp == "OJ"], len[supp == "VC"],
          alternative = "g"))
}

testres[[3]] <- with(tl[[3]],
  t.test(len[supp == "OJ"], len[supp == "VC"],
    alternative = "two"))

```

Now we can arrange test results into a table and determine the hypothesis status

```

## setting up a table for results
results <- setNames(data.frame(matrix(ncol = 7, nrow = 0)),
  c("supp1", "supp2", "dose", "confint1",
    "confint2", "p-value", "status"))

## creating a function to determine whether null hypothesis is rejected based on test results based on v
hypo <- function (t) {
  outZero <- sign(prod(t$conf.int))
  if ((outZero != 1) || (t$p.value > alpha)) {
    "fail to reject"
  } else {
    "reject"
  }
}

## filling the data
for (i in c(1:3)) {
  results[i,"supp1"] <- "OJ"
  results[i,"supp2"] <- "VC"
  results[i,"dose"] <- tl[[i]]$dose[1]
  results[i,"confint1"] <- testres[[i]]$conf.int[1]
  results[i,"confint2"] <- testres[[i]]$conf.int[2]
  results[i,"p-value"] <- testres[[i]]$p.value
  results[i,"status"] <- hypo(testres[[i]])
}

results

```

```

##  supp1 supp2 dose  confint1 confint2      p-value      status
## 1    OJ    VC  0.5  2.346040      Inf 0.0031793034      reject
## 2    OJ    VC  1.0  3.356158      Inf 0.0005191879      reject
## 3    OJ    VC  2.0 -3.798070  3.63807 0.9638515887 fail to reject

```

Conclusions

As a result of t-tests performed on the data we can conclude that: * For the smaller dosages (0.5, 1 mg) the tooth growing effect of orange juice is greater than that of vitamin C with p-values of NA and NA respectively * for the bigger dosage (2 mg) we failed to reject the hypothesis that both supplements have similar effect, and there's no reason to assume that effects of vitamin C and orange juice vary

Environment data

```
sessionInfo()
```

```

## R version 3.5.1 (2018-07-02)
## Platform: x86_64-w64-mingw32/x64 (64-bit)
## Running under: Windows 10 x64 (build 17134)
##
## Matrix products: default
##
## locale:
## [1] LC_COLLATE=English_United States.1252
## [2] LC_CTYPE=English_United States.1252
## [3] LC_MONETARY=English_United States.1252
## [4] LC_NUMERIC=C
## [5] LC_TIME=English_United States.1252
##
## attached base packages:
## [1] stats      graphics  grDevices  utils      datasets  methods    base
##
## other attached packages:
## [1] ggplot2_3.1.0
##
## loaded via a namespace (and not attached):
## [1] Rcpp_0.12.19      bindr_0.1.1       knitr_1.20        magrittr_1.5
## [5] tidyselect_0.2.5  munsell_0.5.0     colorspace_1.3-2  R6_2.3.0
## [9] rlang_0.3.0.1     stringr_1.3.1     plyr_1.8.4        dplyr_0.7.8
## [13] tools_3.5.1       grid_3.5.1        gtable_0.2.0      withr_2.1.2
## [17] htmltools_0.3.6   assertthat_0.2.0  yaml_2.2.0        lazyeval_0.2.1
## [21] rprojroot_1.3-2   digest_0.6.18     tibble_1.4.2      crayon_1.3.4
## [25] bindrcpp_0.2.2    purrr_0.2.5       codetools_0.2-15  glue_1.3.0
## [29] evaluate_0.12     rmarkdown_1.10    labeling_0.3       stringi_1.2.4
## [33] compiler_3.5.1    pillar_1.3.0      scales_1.0.0      backports_1.1.2
## [37] pkgconfig_2.0.2

```