

# Parameteric Tests: Comparison Analysis

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2024-11-20

This is a notebook created by Aliu Adebiyi, the Founder of Alisonia StatGuide

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## Parametric Tests - Comparision:

- In this notebook, I'll describe basic types of comparison test using sample or mock data
- The tests described are t-test (independent sample and dependent otherwise known as paired/matched sample), and one-way analysis of variance (ANOVA)
- All test's assumptions are assumed to be satisfied or wherein stated otherwise
- This is just a practice. In real-life events, non-parametric tests should be adopted if parametric assumptions fail
- This notebook used a daphnia.csv file

## Libaries used

```
library(readxl)          #for reading excel data
library(tidyverse)       #for data manipulation
```

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v dplyr      1.1.0      v readr      2.1.4
## v forcats    1.0.0      v stringr   1.5.0
## v ggplot2    3.5.1      v tibble    3.2.1
## v lubridate  1.9.2      v tidyr     1.3.0
## v purrr      1.0.1
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()    masks stats::lag()
## i Use the http://conflicted.r-lib.org/ to force all conflicts to become errors
```

```
library(ggplot2)          #for graphics
library(ggpubr)           #for graphics
library(caret)
```

```
## Loading required package: lattice
##
## Attaching package: 'caret'
##
## The following object is masked from 'package:purrr':
##
##     lift
```

```
library(rstatix)
```

```
##
## Attaching package: 'rstatix'
##
## The following object is masked from 'package:stats':
##
##     filter
```

## Exploring the data

```
df <- read.csv("daphnia.csv", header = T)
head(df)
```

	growth_rate <dbl>	water <chr>	detergent <chr>	daphnia <chr>	height_b4_fert <dbl>	height_after_fert <dbl>
1	2.919086	Tyne	BrandA	Clone1	3.258246	3.590127
2	2.492904	Tyne	BrandA	Clone1	3.148798	3.760824
3	3.021804	Tyne	BrandA	Clone1	3.529172	3.710617
4	2.350874	Tyne	BrandA	Clone2	3.210920	3.780146
5	3.148174	Tyne	BrandA	Clone2	3.076212	3.879092
6	4.423853	Tyne	BrandA	Clone2	3.262886	3.407060

6 rows

```
str(df)
```

```
## 'data.frame':   72 obs. of  6 variables:
## $ growth_rate   : num  2.92 2.49 3.02 2.35 3.15 ...
## $ water         : chr  "Tyne" "Tyne" "Tyne" "Tyne" ...
## $ detergent     : chr  "BrandA" "BrandA" "BrandA" "BrandA" ...
## $ daphnia       : chr  "Clone1" "Clone1" "Clone1" "Clone2" ...
## $ height_b4_fert : num  3.26 3.15 3.53 3.21 3.08 ...
## $ height_after_fert: num  3.59 3.76 3.71 3.78 3.88 ...
```

```
#Convert the character columns to factors
df <- df %>% mutate_if(is.character, as.factor)

#Display the levels of each variable
apply(df, function(x) if (is.factor(x)) levels(x) else "Is numeric")
```

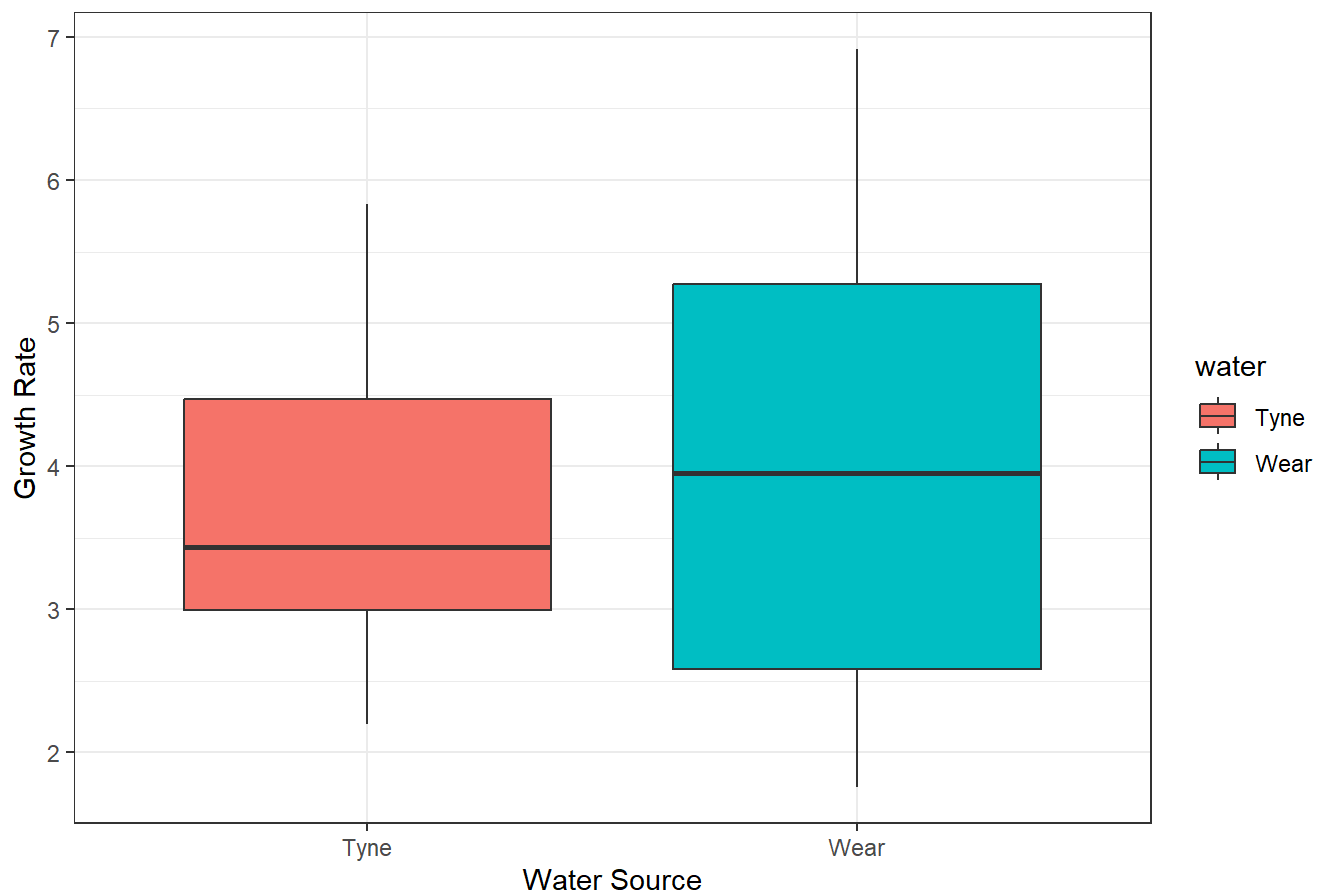
```
## $growth_rate
## [1] "Is numeric"
##
## $water
## [1] "Tyne" "Wear"
##
## $detergent
## [1] "BrandA" "BrandB" "BrandC" "BrandD"
##
## $daphnia
## [1] "Clone1" "Clone2" "Clone3"
##
## $height_b4_fert
## [1] "Is numeric"
##
## $height_after_fert
## [1] "Is numeric"
```

## Two-sample t-test - Independent

- Data types: DV(scale, e.g. *growth\_rate*) IV(two-level categorical variable such as *water -Tyne/Wear*)
- Independent of observations, that is, level came from different groups

```
#1. Exploring the target variables
ggplot(df, aes(x = water, y = growth_rate, fill = water)) +
  geom_boxplot() +
  labs(title = "Boxplot of Growth Rate by Water Source", x = "Water Source", y = "Growth Rate")
+
  theme_bw()
```

Boxplot of Growth Rate by Water Source



### ### 2. Summary Statistics of Growth Rate by Water

```
summary_stats <- df %>%  
  group_by(water) %>%  
  summarise(  
    Mean = mean(growth_rate),  
    Median = median(growth_rate),  
    SD = sd(growth_rate),  
    Min = min(growth_rate),  
    Max = max(growth_rate),  
    .groups = 'drop'  
  )  
print(summary_stats)
```

```
## # A tibble: 2 x 6  
##   water Mean Median   SD   Min   Max  
##   <fct> <dbl> <dbl> <dbl> <dbl> <dbl>  
## 1 Tyne   3.69   3.44  1.05  2.20  5.83  
## 2 Wear   4.02   3.96  1.48  1.76  6.92
```

### #3. Perform the t-test

```
t_test <- t.test(growth_rate ~ water, data = df, var.equal = TRUE)  
print(t_test)
```

```
##
## Two Sample t-test
##
## data: growth_rate by water
## t = -1.0984, df = 70, p-value = 0.2758
## alternative hypothesis: true difference in means between group Tyne and group Wear is not equal to 0
## 95 percent confidence interval:
## -0.9350744 0.2709019
## sample estimates:
## mean in group Tyne mean in group Wear
## 3.685862 4.017948
```

## REPORT

To examine whether there is a significant difference in growth rates between the two water sources, Tyne and Wear, an independent samples t-test was conducted. The analysis compared the mean growth rates of the two groups.

The results of the t-test indicated that there was no statistically significant difference in the growth rates between the two water sources,  $t(70) = -1.10$ ,  $p = .276$ . The mean growth rate for the Tyne group ( $M = 3.69$ ) was slightly lower than that of the Wear group ( $M = 4.02$ ). However, the 95% confidence interval for the difference in means  $[-0.94, 0.27]$  includes zero, further suggesting that there is no significant difference.

In conclusion, the data do not provide evidence to support a difference in growth rates between water sources Tyne and Wear.

# Two-sample t-test - Dependent

- Data types: DV(scale, e.g. *height*) IV(two-level categorical variable such as *before/after*)
- Dependent observations, that is, levels came from same group

```
#1. Summary statistics
summary_stats <- df %>%
  summarise(
    mean_before = mean(height_b4_fert, na.rm = TRUE),
    sd_before = sd(height_b4_fert, na.rm = TRUE),
    mean_after = mean(height_after_fert, na.rm = TRUE),
    sd_after = sd(height_after_fert, na.rm = TRUE),
    n = n()
  )

print(summary_stats)
```

```
## mean_before sd_before mean_after sd_after n
## 1 3.446607 0.2933281 3.676593 0.236158 72
```

```
# 2. Paired t-test
t_test_result <- t.test(df$height_b4_fert, df$height_after_fert, paired = TRUE)

# Print t-test results
print(t_test_result)
```

```
##
## Paired t-test
##
## data: df$height_b4_fert and df$height_after_fert
## t = -5.1677, df = 71, p-value = 2.086e-06
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.3187254 -0.1412474
## sample estimates:
## mean of the differences
## -0.2299864
```

## REPORT

A paired-samples t-test was conducted to compare the mean heights of plants before fertilizer application and after fertilizer application. The results indicated a statistically significant difference between the two conditions,  $t(71) = -5.168$ ,  $p < .001$ . The mean difference in height ( $M = -0.230$ , 95% CI: [-0.319, -0.141]) suggests that plants were, on average, shorter before the application of fertilizer compared to after the application.

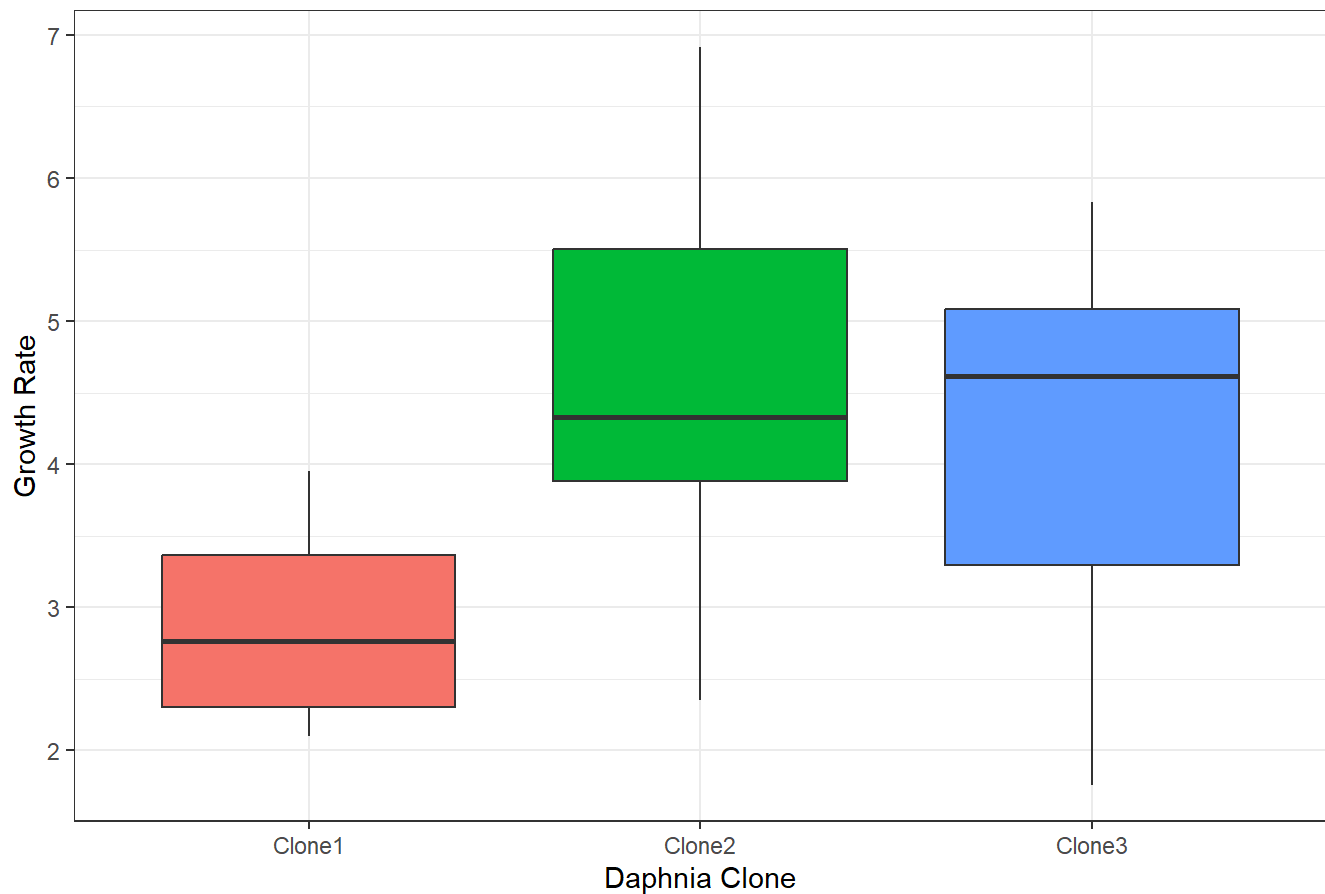
This finding indicates that fertilizer application had a significant effect on plant height, with plants growing taller after receiving fertilizer. The negative mean difference reflects the increase in plant height post-application. These results support the hypothesis that fertilizer application enhances plant growth.

# One-way Analysis of Variance

- Data types: DV(scale, e.g. *growth\_rate*) IV(three or more levels categorical variable such as *daphnia*)
- Independent observations, that is, levels came from different groups
- If levels come from the same group, run Repeated measures ANOVA (not cover here)

```
#1. Explore the target variables
ggplot(df, aes(x = daphnia, y = growth_rate, fill = daphnia)) +
  geom_boxplot() +
  labs(title = "Boxplot of Growth Rate by Daphnia Group",
       x = "Daphnia Clone",
       y = "Growth Rate") +
  theme_bw() +
  theme(legend.position = "none")
```

# Boxplot of Growth Rate by Daphnia Group



## #2. Summary statistics

```
summary_stats <- df %>%
  group_by(daphnia) %>%
  summarise(
    Mean = mean(growth_rate, na.rm = TRUE),
    SD = sd(growth_rate, na.rm = TRUE),
    Median = median(growth_rate, na.rm = TRUE),
    Min = min(growth_rate, na.rm = TRUE),
    Max = max(growth_rate, na.rm = TRUE)
  )

print(summary_stats)
```

```
## # A tibble: 3 x 6
##   daphnia Mean    SD Median  Min   Max
##   <fct>   <dbl> <dbl>   <dbl> <dbl> <dbl>
## 1 Clone1  2.84 0.576  2.76  2.10  3.95
## 2 Clone2  4.58 1.24   4.33  2.35  6.92
## 3 Clone3  4.14 1.24   4.62  1.76  5.83
```

```
#3 Run the test ANOVA
anova_result <- aov(growth_rate ~ daphnia, data = df)

# ANOVA summary
summary(anova_result)
```

```
##          Df Sum Sq Mean Sq F value    Pr(>F)
## daphnia     2   39.18    19.59    17.33 7.96e-07 ***
## Residuals   69   77.98     1.13
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

## RESULT

A one-way ANOVA was conducted to compare the effect of different *Daphnia* clones on plant growth rate. The analysis revealed a statistically significant effect of *Daphnia* clone on growth rate,  $F(2,69)=17.33$ ,  $p<.001$ . This indicates that the growth rate differs significantly across the three *Daphnia* clones.

Since the overall test was significant, pairwise comparisons are necessary to identify which specific groups differ from one another. The boxplot below visually represents the growth rates for each *Daphnia* group, with pairwise comparisons of the differences in means included. These comparisons provide further insight into how growth rates vary between the clones.

```
res.aov <- anova_test(data = df, dv = growth_rate, between = daphnia)
byp_anova <- ggboxplot(df, x = "daphnia", y = "growth_rate") + theme_bw()

pwc_anova <- df %>%
  pairwise_t_test(
    growth_rate ~ daphnia, paired = FALSE,
    p.adjust.method = "bonferroni"
  )
pwc_anova
```

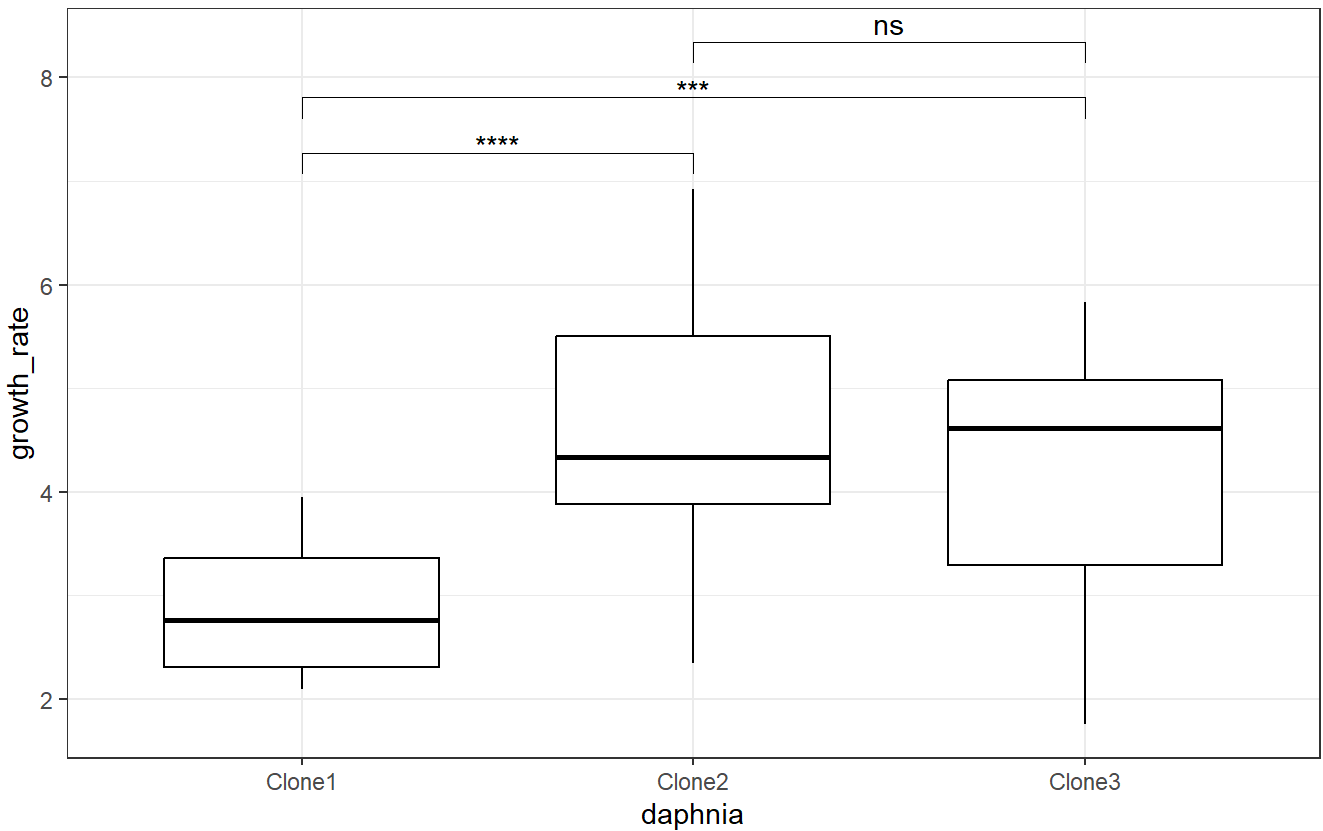
.y. <chr>	group1 <chr>	group2 <chr>	n1 <int>	n2 <int>	p <dbl>	p.signif <chr>	p.adj <dbl>	p.adj.signif <chr>
1 growth_rate	Clone1	Clone2	24	24	3.18e-07	****	9.54e-07	****
2 growth_rate	Clone1	Clone3	24	24	6.99e-05	****	2.10e-04	***
3 growth_rate	Clone2	Clone3	24	24	1.58e-01	ns	4.73e-01	ns



```
pwc_anova <- pwc_anova %>% add_xy_position(x = "daphnia")

bxp_anova +
  stat_pvalue_manual(pwc_anova) +
  labs(
    subtitle = get_test_label(res.aov, detailed = TRUE),
    caption = get_pwc_label(pwc_anova)
  )
```

Anova,  $F(2,69) = 17.33$ ,  $p = <0.0001$ ,  $\eta_g^2 = 0.33$



pwc: **T test**; p.adjust: **Bonferroni**

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