

# Collaborative 3DUI Data Analysis

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## Data analysis

This is a on going analysis of the data collected in the user experiment performed...

## Summary

First we set the environment, load and show the raw data.

```
setwd("/home/jeronimo/Documents/DataAnalysis3DController/Analysis/R")
dAvgPerTask <- read.csv("SummaryToR.csv")
dAvgPerTask
```

##	Members	T1	T2	T3	T4	E1	E2
## 1	1	44.60100	27.73000	44.5060	56.69900	4.280380	16.953010
## 2	1	86.26980	26.60070	129.5357	88.65650	3.869670	6.642847
## 3	1	37.22350	32.50000	130.7602	63.44300	1.317529	9.220284
## 4	1	71.42880	55.88850	281.5956	225.20610	4.024921	8.576299
## 5	1	32.65650	41.49640	105.6555	139.63410	4.456124	9.002918
## 6	2	19.92100	23.75830	86.0130	90.59442	1.923633	11.753480
## 7	2	13.64980	15.95330	87.2619	113.11968	3.004579	4.069859
## 8	2	16.45030	14.49990	69.3649	58.41970	2.192855	5.547572
## 9	2	17.20950	29.62760	139.3775	111.04040	2.576868	8.645294
## 10	2	59.33450	16.48410	109.5334	83.16500	4.601643	8.891509
## 11	2	35.49839	34.55538	99.4809	93.11320	1.390942	4.142436
## 12	3	58.66863	16.39880	127.2309	82.56380	0.867342	3.243309
## 13	3	29.00930	49.16030	120.6714	91.11170	2.359320	4.060901
## 14	3	36.21300	22.41350	76.8215	55.29020	2.259253	3.317118
## 15	3	21.44110	17.18680	86.8132	61.24430	1.300738	3.357939
## 16	3	20.20286	25.70100	283.4138	192.72400	2.329338	4.309965
## 17	3	24.40880	42.26010	41.2966	38.11630	2.187007	8.903512
## 18	4	47.19800	21.40800	162.9620	87.06737	1.554914	3.538198
## 19	4	22.00295	16.25100	74.4534	71.80547	1.402203	4.501410
## 20	4	21.09751	16.95763	56.4048	63.90760	3.145458	6.065219
## 21	4	54.20073	37.12230	127.1752	60.76750	1.623129	2.768127
##	E3	E4					
## 1	9.070060	8.529819					
## 2	7.381541	5.705525					
## 3	6.207668	5.088869					
## 4	3.341505	3.299659					
## 5	6.817599	8.455383					
## 6	3.251326	3.595758					
## 7	4.849439	4.512303					
## 8	6.037892	4.339551					
## 9	4.726371	3.707839					
## 10	4.380987	5.540681					
## 11	4.699787	4.160006					
## 12	2.972291	3.701374					

```
## 13 3.978512 3.053629
## 14 4.150805 4.628316
## 15 3.179116 4.247869
## 16 2.652203 2.953497
## 17 4.778647 4.583976
## 18 2.389965 2.110792
## 19 3.337539 2.288988
## 20 3.023767 3.257560
## 21 1.776181 1.903549
```

The data is arranged by Task vs. Team members. Columns are organized as follows:

- Members: Number of users in the team;
- T1, T2, T3, T4: Time to complete task 1 to 4;
- E1, E2, E3, E4: Errors in task 1 to 4.

We split the data for better manipulation:

```
dAvgTime <- subset(dAvgPerTask, select = c(2,3,4,5))
dAvgError <- subset(dAvgPerTask, select = c(6,7,8,9))
```

Below the data is summarized:

```
describe(dAvgTime)
```

```
##      vars  n   mean    sd median trimmed  mad   min    max  range skew
## T1     1 21  36.60 19.92  32.66   34.17 18.46 13.65  86.27  72.62 0.87
## T2     2 21  27.81 11.90  25.70   26.38 13.13 14.50  55.89  41.39 0.77
## T3     3 21 116.21 63.86 105.66  105.27 37.22 41.30 283.41 242.12 1.46
## T4     4 21  91.79 45.68  83.17   83.31 32.50 38.12 225.21 187.09 1.57
##      kurtosis    se
## T1    -0.25   4.35
## T2    -0.50   2.60
## T3     1.64 13.93
## T4     1.84   9.97
```

```
describe(dAvgError)
```

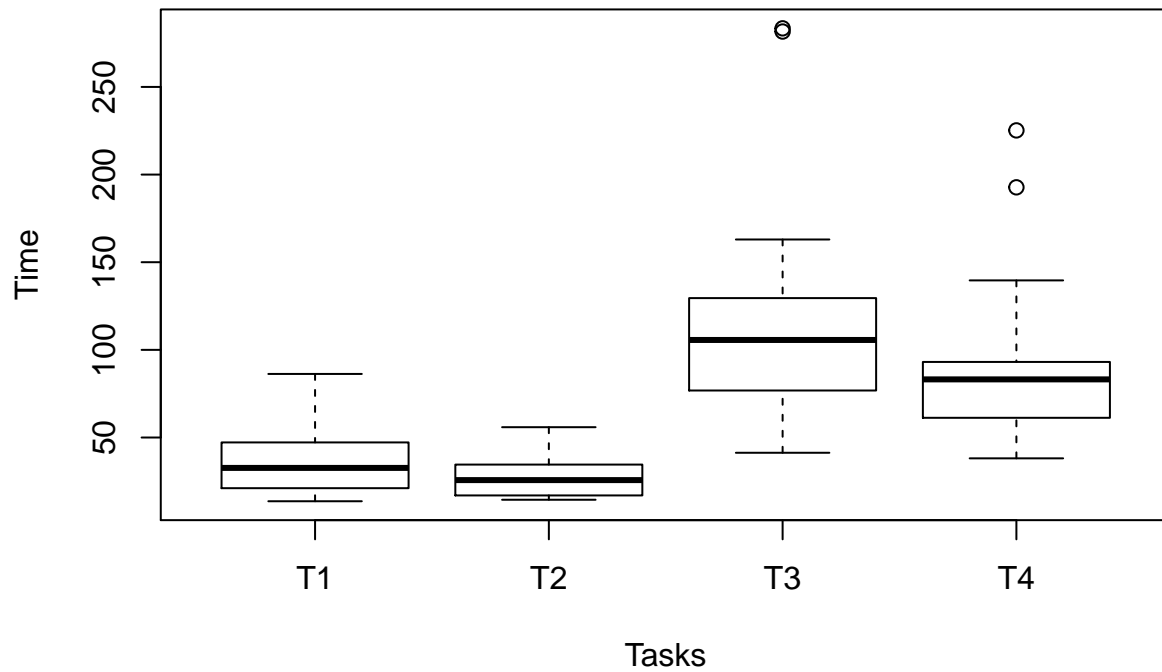
```
##      vars  n mean    sd median trimmed  mad   min    max range skew kurtosis
## E1     1 21 2.51 1.15   2.26   2.44 1.27 0.87  4.60  3.73 0.50   -1.15
## E2     2 21 6.55 3.55   5.55   6.05 3.31 2.77 16.95 14.18 1.17    1.04
## E3     3 21 4.43 1.82   4.15   4.26 1.44 1.78  9.07  7.29 0.83   -0.04
## E4     4 21 4.27 1.74   4.16   4.04 1.34 1.90  8.53  6.63 1.07    0.71
##      se
## E1 0.25
## E2 0.77
## E3 0.40
## E4 0.38
```

## Plots

Time of task completion vs. Task for all combinations of teams: The code used to generate the charts is:

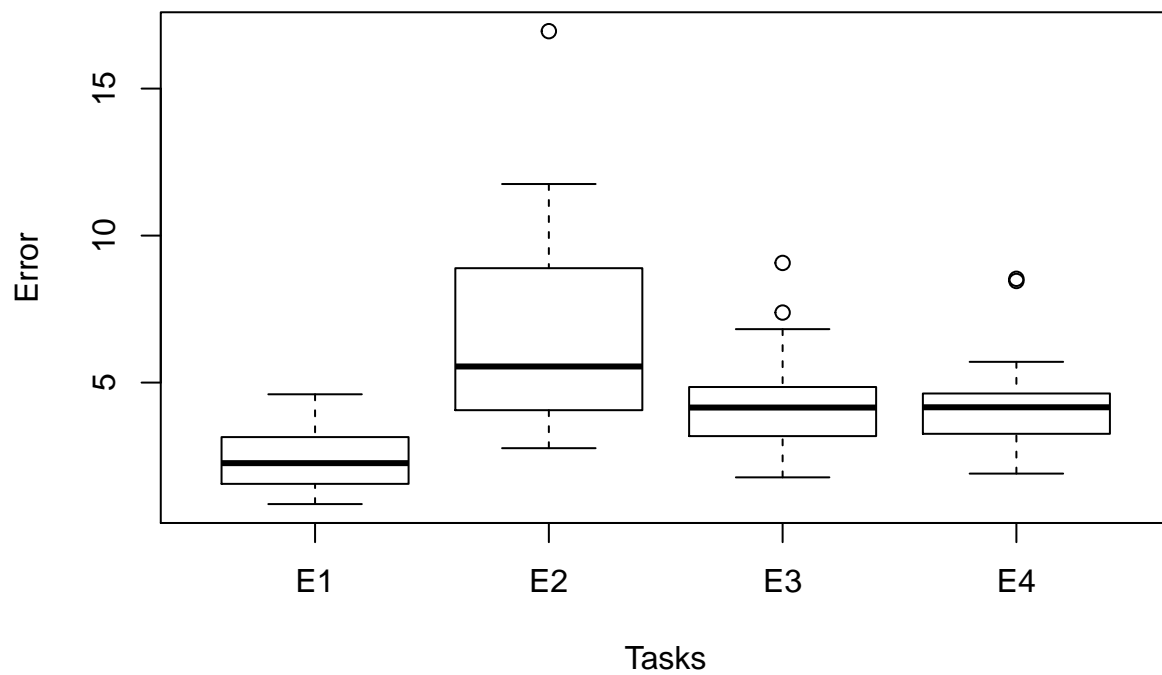
```
boxplot(dAvgTime,xlab="Tasks",ylab="Time",main="Time to complete the tasks")
```

## Time to complete the tasks



```
boxplot(dAvgError,xlab="Tasks",ylab="Error",main="Error perfered in the tasks")
```

## Error perfered in the tasks



## Analysis of time of completion per task

### Shapiro

First, we perform the Shapiro normality test. This test determine if the data is normally distributed. It is important to determine if the data is normally distributed to conduce posterior tests.

```
shap_dT <- lapply(dAvgTime, shapiro.test)
shap_dE <- lapply(dAvgError, shapiro.test)
res_shap_dT <- sapply(shap_dT, `[,`, c("statistic", "p.value"))
res_shap_dE <- sapply(shap_dE, `[,`, c("statistic", "p.value"))
```

```
res_shap_dT
```

```
##           T1           T2           T3           T4
## statistic 0.8974621 0.9017898 0.812624 0.8037059
## p.value   0.03125427 0.0379083 0.001029792 0.0007478811
```

```
res_shap_dE
```

```
##           E1           E2           E3           E4
## statistic 0.9139373 0.8525865 0.9313148 0.8860365
## p.value   0.06570504 0.004708184 0.1462635 0.0189353
```

As we can see, the  $p$  - *value* of most Shapiro tests reveled that the data are not normally distributed. Since in this test the comparisons are made with the same subjects and we are varying the tasks, the next step is to perform a Friedman analysis.

### Friedman

Friedman test is a non-parametric randomized block analysis of variance. Which is to say it is a non-parametric version of a one way ANOVA with repeated measures. That means that while a simple ANOVA test requires the assumptions of a normal distribution and equal variances (of the residuals), the Friedman test is free from those restriction. The price of this parametric freedom is the loss of power (of Friedman's test compared to the parametric ANOVA versions).

The hypotheses for the comparison across repeated measures are:

- H0: The distributions (whatever they are) are the same across repeated measures
- H1: The distributions across repeated measures are different

The test statistic for the Friedman's test is a Chi-square with [(number of repeated measures)-1] degrees of freedom. A detailed explanation of the method for computing the Friedman test is available on Wikipedia.