

# Assignment-45488479

2022-10-28

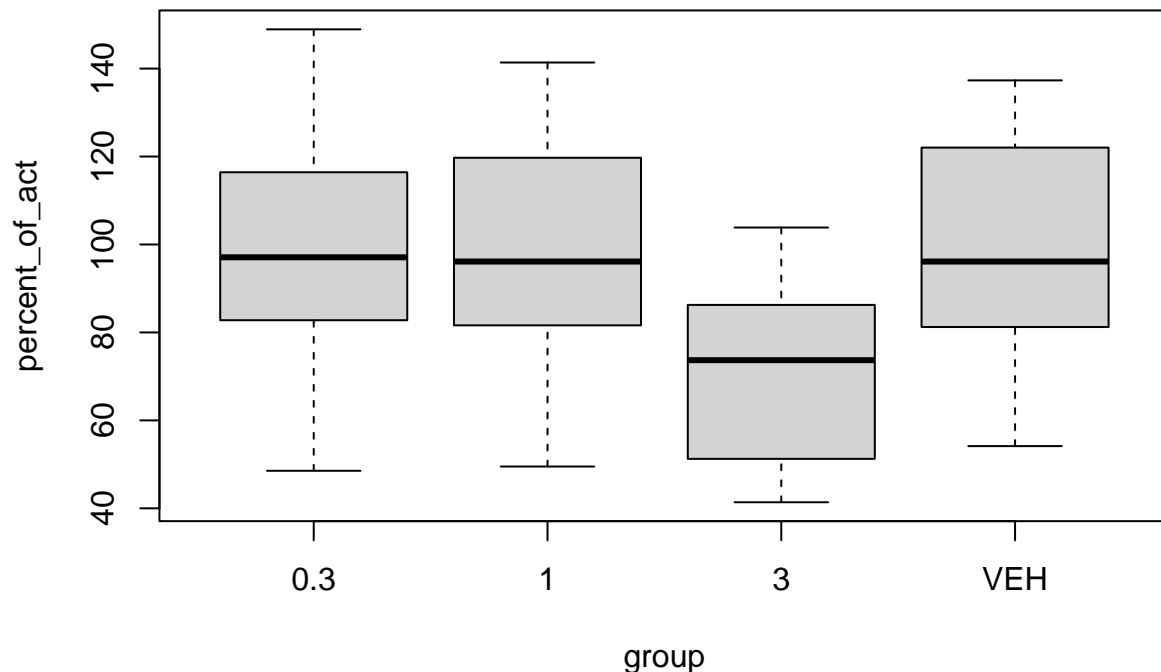
```
library("tinytex")
library("tidyverse")
```

```
## -- Attaching packages ----- tidyverse 1.3.2 --
## v ggplot2 3.3.6      v purrr  0.3.4
## v tibble  3.1.8      v dplyr  1.0.9
## v tidyr   1.2.0      v stringr 1.4.0
## v readr   2.1.2      v forcats 0.5.1
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()    masks stats::lag()
```

```
mice <- read.table("mice_pot.txt", header = TRUE)
head(mice)
```

```
##   group percent_of_act
## 1   0.3      98.82671
## 2   0.3     116.42599
## 3   0.3     132.09128
## 4   0.3      84.12842
## 5   0.3     148.91697
## 6   0.3      48.54306
```

```
boxplot(percent_of_act ~ group, data = mice)
```



#a. ##The variances look larger in the 0.3mg/kg plot than 3mg/kg. ##The boxes appear to be equally sized implying that the different formations appear to have the same variability. ## Including Plots

You can also embed plots, for example:

```
mice.aov = aov(percent_of_act ~ group, data = mice)
anova(mice.aov)
```

```
## Analysis of Variance Table
##
## Response: percent_of_act
##           Df Sum Sq Mean Sq F value Pr(>F)
## group      3   6329  2109.65   3.1261 0.0357 *
## Residuals 42  28344   674.85
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

##b. ## • Hypotheses:  $H_0: \mu_1 = \mu_2 = \mu_3 = \mu_4$  (no effect in the population);  $H_1$ : not all means are equal (an effect in the population). ## • Test statistic:  $F_{obs} = 3.1261$  ## • Null distribution: If  $H_0$  is true,  $F_{obs}$  behaves like a  $F_{3,42}$  distribution ## • P-Value:  $P(F_{3,42} \geq 3.1261) = 0.0357 < 0.05$  ## • Conclusion: Since the P-Value is less than the significance level of 0.05 (5%) we have evidence to reject  $H_0$  in favour of  $H_1$ . That is, we have evidence that the mice content is not all the same for the different formations.

Note that the `echo = FALSE` parameter was added to the code chunk to prevent printing of the R code that generated the plot.

```
TukeyHSD(mice.aov)
```

```
## Tukey multiple comparisons of means
## 95% family-wise confidence level
##
## Fit: aov(formula = percent_of_act ~ group, data = mice)
##
## $group
##          diff          lwr          upr      p adj
## 1-0.3      1.7298436 -28.9122207 32.371908 0.9987527
## 3-0.3     -26.6546330 -58.5829330  5.273667 0.1309483
## VEH-0.3    2.6774970 -26.6219205 31.976915 0.9947868
## 3-1       -28.3844765 -58.1381757  1.369223 0.0663434
## VEH-1      0.9476534 -25.9655801 27.860887 0.9996956
## VEH-3     29.3321300   0.9630909 57.701169 0.0403151
```

##c. ##t-test value is 3.1262. and associated degrees of freedom is F3,42. ##Although P-value = 0.0357 < 0.05, the H0 is rejected.

```
kml <- read.table("kml.dat", header = TRUE)
head(kml)
```

```
##          kmL driver  car
## 1 10.75614      A  one
## 2 10.71363      A  one
## 3 12.28666      A  two
## 4 12.75432      A  two
## 5 10.54357      A three
## 6 10.67111      A three
```

```
summary(aov(kmL ~ driver * car, data = kml))
```

```
##          Df Sum Sq Mean Sq F value    Pr(>F)
## driver      3  50.66  16.887   531.60 < 2e-16 ***
## car          4   17.12   4.280   134.73 3.66e-14 ***
## driver:car  12    0.44   0.037    1.16   0.371
## Residuals   20    0.64   0.032
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
summary(aov(kmL ~ car * driver, data = kml))
```

```
##          Df Sum Sq Mean Sq F value    Pr(>F)
## car          4   17.12   4.280   134.73 3.66e-14 ***
## driver      3  50.66  16.887   531.60 < 2e-16 ***
## car:driver  12    0.44   0.037    1.16   0.371
## Residuals   20    0.64   0.032
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
checkbalance = kml; checkbalance[15, ] = c(90, "truck", "km")

summary(aov(kmL ~ driver * car, data=checkbalance))
```

```
##           Df Sum Sq Mean Sq  F value    Pr(>F)
## driver      4   5869    1467 44128.610 < 2e-16 ***
## car         4     16       4   119.828 3.37e-13 ***
## driver:car  12      0       0    1.061   0.44
## Residuals   19      1       0
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

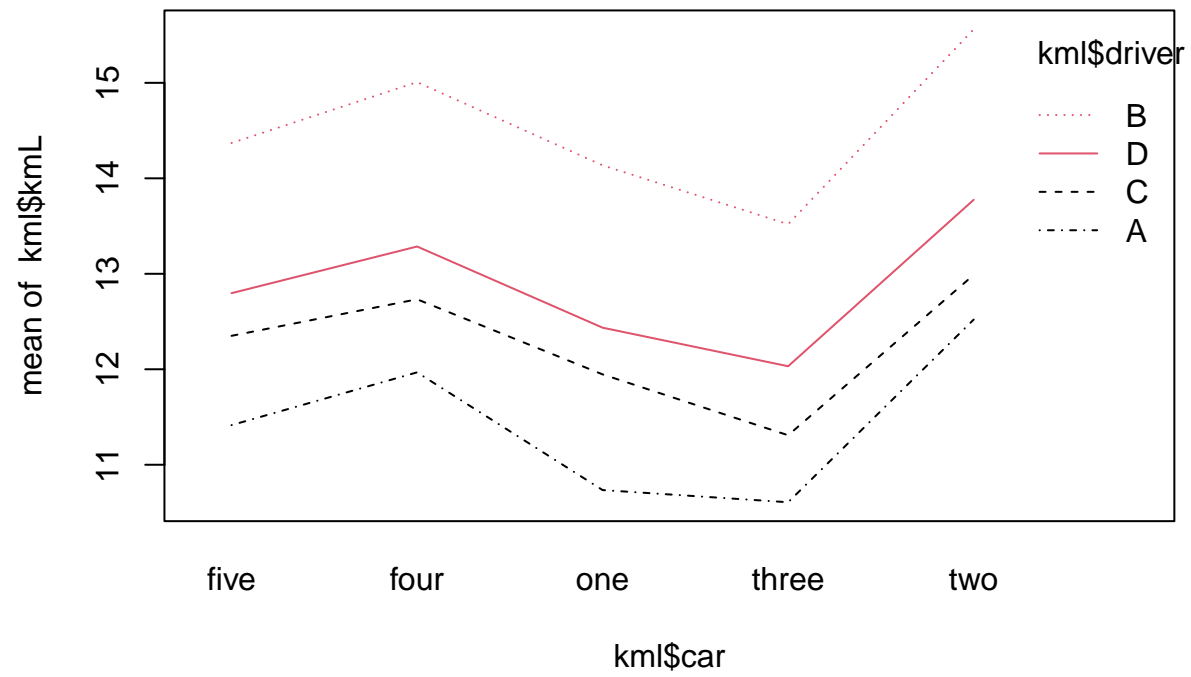
```
summary(aov(kmL ~ car * driver, data=checkbalance))
```

```
##           Df Sum Sq Mean Sq  F value Pr(>F)
## car         5   5837  1167.4 35111.696 <2e-16 ***
## driver      3     48    15.9   478.424 <2e-16 ***
## car:driver  12      0     0.0    1.061   0.44
## Residuals   19      1     0.0
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

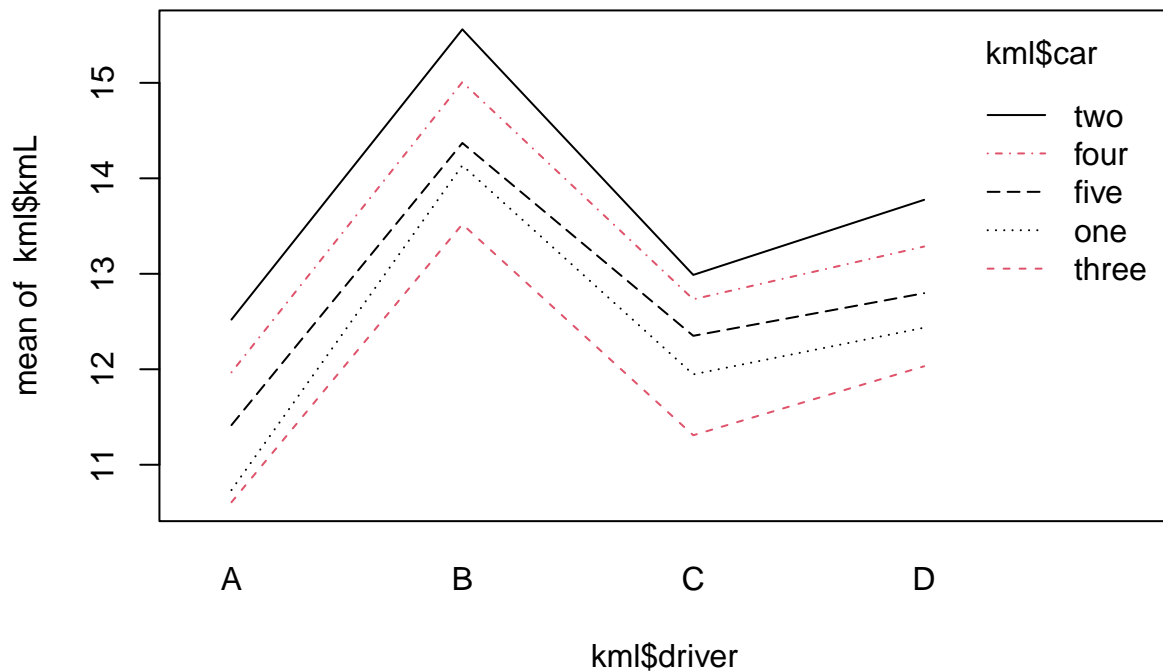
##Q2 ##a. Yes the design is balanced since there are two entries for a combination of each of the Car and driver

##b.

```
interaction.plot(kml$car, kml$driver, kml$kmL, col = 1:2)
```



```
interaction.plot(kml$driver, kml$car, kml$kmL, col = 1:2)
```



```
kml1 = lm(kmL ~ driver, data = kml)
anova(kml1)
```

```
## Analysis of Variance Table
##
## Response: kmL
##          Df Sum Sq Mean Sq F value    Pr(>F)
## driver     3 50.661  16.8869   33.409 1.672e-10 ***
## Residuals  36  18.197    0.5055
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
kml2 = lm(kmL ~ car, data = kml)
anova(kml2)
```

```
## Analysis of Variance Table
##
## Response: kmL
##          Df Sum Sq Mean Sq F value    Pr(>F)
## car        4  17.119   4.2798   2.8952 0.03597 *
## Residuals  35  51.738   1.4782
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

##c. ##Null hypothesis : There is no Dependence of drivers. ##Alternative hypothesis : There is some

dependence on drivers and fuel efficiencies. ##From kml1, t-stat is 33.409, P-value is 1.672e-10. ##Since P-value is less than 0.05, null hypothesis is rejected.

##Null hypothesis: There is no dependence on cars and fuel efficiencies. ##Alternate hypothesis: There is some dependence on cars and fuel efficiencies. ##From kml2, t-stat is 2.8952, P-value is 0.03597. ##Since P-value is less than 0.05, null hypothesis is rejected.

##d ##In both the Cases null hypothesis is rejected (See Q2-c). ##This implies that Cars and drivers on fuel efficiencies choice of in Kml.