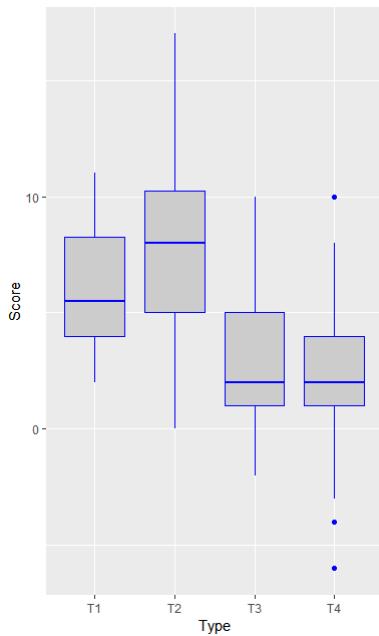


### Ex 8.3.1

```
> hw831 <- read.csv("C:/Users/Teaching/ma4310/EXR_C08_S03_01.csv")
> hw831$type = factor(hw831$type, labels = c("T1", "T2", "T3", "T4"))
> hw831$subject = factor(hw831$subject, levels=unique(hw831$subject))
> require(ggplot2)
> require(ggplot2)
> ggplot(hw831, aes(x = type, y = score)) +
+     geom_boxplot(fill = "grey80", colour = "blue") +
+     scale_x_discrete() + xlab("Type") +
+     ylab("Score")
```



```
> hw831.model = lm(score~subject+type, data = hw831)
> anova(hw831.model)
```

Two-way anova - additive model

Analysis of Variance Table

```
Response: score
          Df Sum Sq Mean Sq F value    Pr(>F)
subject    23 481.12 20.918  2.4923 0.001887 ***
type       3 498.37 166.125 19.7930 2.306e-09 ***
Residuals 69 579.12   8.393
---
```

```
> qf(0.95,df1=3,df2=69)
[1] 2.737492
```

**Conclusion:**  $F = 19.79 > 2.737492$ . Also  $p < 0.001$ . Hence reject  $H_0$ . Improvement scores differed significantly among the four training types.

### Ex 8.3.2

```
> # NOTE: I MANUALLY MODIFIED DATA INPUT FILE TO FIX FORMAT
> hw832 <- read.csv("C:/Users/Teaching/ma4310/EXR_C08_S03_02b.csv")
> hw832$type = factor(hw832$type, levels=unique(hw832$type))
> hw832$subject = factor(hw832$subject, levels=unique(hw832$subject))
> hw832.model = lm(score~subject+type, data = hw832)
> anova(hw832.model)
```

Two-way anova - additive model

Analysis of Variance Table

Response: score

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
subject	9	1694868	188319	24.1161	1.121e-10 ***
type	3	153813	51271	6.5658	0.001775 **
Residuals	27	210839	7809		

**Conclusion:** F = 6.57, p = 0.002. Reject H0. Tic frequencies differed significantly among the time periods

### Ex 8.4.1

```
> hw841 <- read.csv("C:/Users/Teaching/ma4310/EXR_C08_S04_01.csv")
> hw841$time = factor(hw841$time, levels=unique(hw841$time))
> hw841$subj = factor(hw841$subj, levels=unique(hw841$subj))
> hw841.model = lm(outcome~subj+time, data = hw841)
> anova(hw841.model)
```

Two-way anova - additive model

Analysis of Variance Table

Response: outcome

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
subj	9	64.225	7.136	5.4313	0.0002846 ***
time	3	192.275	64.092	48.7801	4.939e-11 ***
Residuals	27	35.475	1.314		

**Conclusion:** F = 48.78, p < 0.001. Reject H0. There was significant difference in the four time periods

### Ex 8.4.2

```
> hw842 <- read.csv("C:/Users/Teaching/ma4310/EXR_C08_S04_02.csv")
> hw842$LOAD = factor(hw842$LOAD, levels=unique(hw842$LOAD))
> hw842$SPECIMEN = factor(hw842$SPECIMEN, levels=unique(hw842$SPECIMEN))
> hw842.model = lm(LAXITY~SPECIMEN+LOAD, data = hw842)
> anova(hw842.model)
```

Two-way anova – additive model

Analysis of Variance Table

Response: LAXITY

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
SPECIMEN	6	195951	32659	66.5885	8.96e-14 ***
LOAD	4	14892	3723	7.5909	0.0004221 ***
Residuals	24	11771	490		

**Conclusion:**  $F = 7.59$ ,  $p < 0.001$ . Reject  $H_0$ . Graft laxity was significantly different among the five load levels

### Ex 8.5.2

```
> hw852 <- read.csv("C:/Users/Teaching/ma4310/EXR_C08_S05_02.csv")
> hw852$PSYCH = factor(hw852$PSYCH, labels = c("T1", "T2", "T3", "T4"))
> hw852$PHYS = factor(hw852$PHYS)
> hw852.model = lm(MONTHS~PHYS+PSYCH+PHYS:PSYCH, data = hw852)
> anova(hw852.model)
```

Two-way anova – additive model with interaction

Analysis of Variance Table

Response: MONTHS

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
PHYS	5	13.796	2.7592	7.5913	2.552e-05 ***
PSYCH	3	90.408	30.1359	82.9112	< 2.2e-16 ***
PHYS:PSYCH	15	32.001	2.1334	5.8694	1.203e-06 ***
Residuals	48	17.447	0.3635		

**Conclusion:** Since the  $p$  values are significant for physical therapy treatment, psychological treatment and interaction, we reject the null hypothesis for each of those variables.

**QUIZ #4** A study investigated the awareness of deficit profiles among stroke patients undergoing rehabilitation. The study included 35 patients with a stroke lesion in the **right hemisphere** and 19 patients with a lesion on the **left hemisphere**. In addition, lesion size was grouped as **size-2**: 1-3 cm; **size-3**: 3-5 cm”; and **size-4**: 5 cm or greater. The outcome variables is a measure of each patient’s total unawareness of their own limitations, with scores ranging from 8 to 24 and higher scores indicating more unawareness.

- (a) Explore the data and make sure each factor is identified correctly. Is the experiment design balanced or unbalanced? Justify your conclusion.
- (b) Apply the correct Anova method to test for a difference in lesion size, hemisphere, and interaction, using significance level 0.05 for all tests. Discuss the conclusion of the test.
- (c) Use the Tukey’s HSD procedure to test for significant differences among individual pairs of means, if appropriate. Justify your conclusion.

```
> dataq4 <- read.csv("C:/Users/dlabate/Desktop/Teaching/ma4310/dataq4.csv")
> str(dataq4)
'data.frame': 54 obs. of 3 variables:
 $ Score: int 11 13 10 11 9 10 9 8 10 13 ...
 $ Side : chr "L" "L" "L" "L" ...
 $ Size : int 2 2 2 2 2 2 2 2 2 3 ...
> dataq4$Side <- factor(dataq4$Side, levels = c("L","R"),labels = c("L","R"))
> dataq4$Size <- factor(dataq4$Size, levels = c(2,3,4),labels = c("2","3","4"))
> str(dataq4)
'data.frame': 54 obs. of 3 variables:
 $ Score: int 11 13 10 11 9 10 9 8 10 13 ...
 $ Side : Factor w/ 2 levels "L","R": 1 1 1 1 1 1 1 1 1 1 ...
 $ Size : Factor w/ 3 levels "2","3","4": 1 1 1 1 1 1 1 1 1 2 ...
113 obs. of 2 variables:
```

a) Analysis of variance

```
> table(dataq4$Side, dataq4$Size)

 2 3 4
L 9 5 5
R 14 9 12
```

The table shows that the experiment has unequal numbers of subjects in each group.  
Hence, we need to run the Anova test with **Unbalanced Design**.

b) We apply the analysis of variance hypothesis testing

```
> res.aov <- aov(Score ~ Side*Size, data = dataq4)
> library(car)
> Anova(res.aov, type = "III")
```

Anova Table (Type III tests)

Response: Score					
	Sum Sq	Df	F value	Pr(>F)	
(Intercept)	920.11	1	227.8446	< 2e-16	***
Side	0.18	1	0.0452	0.83253	
Size	30.10	2	3.7269	0.03129	*
Side:Size	10.54	2	1.3050	0.28062	
Residuals	193.84	48			

Conclusion: since p-value < 0.05 for the size factor, the **lesion size is a significant factor** in the awareness deficit profiles.  
On the other hand, **side and interaction are not statistically significant factors**.

NOTE: if we do not apply the Type III test, we obtain a different result

```
> summary(res.aov)
```

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
Side	1	1.29	1.287	0.319	0.5750
Size	2	29.98	14.991	3.712	0.0317 *
Side:Size	2	10.54	5.270	1.305	0.2806
Residuals	48	193.84	4.038		

c) We apply the Tukey's HSD procedure

```
> TukeyHSD(res.aov, which = "Size")

Tukey multiple comparisons of means
 95% family-wise confidence level

Fit: aov(formula = Score ~ Side * Size, data = dataq4)

$Size
    diff      lwr      upr     p adj
3-2 0.5824733 -1.0649968 2.229943 0.6709014
4-2 1.7373036  0.1828198 3.291787 0.0252228
4-3 1.1548304 -0.5991994 2.908860 0.2587441
```

Conclusion: based on the values of p adj, only the difference between the sizes 2 and 4 is statistically significant.