

Exercises 7 - SOLUTIONS

1. Using the data frame `keyboard` [see solutions to Exercises 6]. Note: you may also need to refit the ANOVA object `keyboard.aov`.

(a) `> pairwise.t.test(score, method, p.adj="none")`

Pairwise comparisons using t tests with pooled SD

data: score and method

	A	B	C
B	0.03671	-	-
C	0.01047	0.59845	-
D	0.09277	0.00041	8.5e-05

P value adjustment method: none

```
> keyboard.aov <- aov(score ~ method, data = keyboard)
> Tukey.keyboard <- TukeyHSD(keyboard.aov)
> Tukey.keyboard
  Tukey multiple comparisons of means
    95% family-wise confidence level
```

Fit: aov(formula = score ~ method, data = keyboard)

\$method	diff	lwr	upr	p adj
B-A	-4.9	-10.982495	1.18249453	0.1512083
C-A	-6.1	-12.182495	-0.01750547	0.0491186
D-A	3.9	-2.182495	9.98249453	0.3250603
C-B	-1.2	-7.282495	4.88249453	0.9508776
D-B	8.8	2.717505	14.88249453	0.0022085
D-C	10.0	3.917505	16.08249453	0.0004749

The Fisher approach shows that the effects of Methods A and D are both significantly better than the effects of both of Methods B and C. The effects of Methods A and D are not significantly different from each other. The effects of Methods B and C are not significantly different from each other.

The conclusions from the more conservative Tukey approach are essentially the same, except that the effects of Methods A and B do not differ significantly from each other.

```
(b) > c1 <- c(3, -1, -1, -1)
> c2 <- c(0, 2, -1, -1)
> c3 <- c(0, 0, 1, -1)
>
> ctr <- matrix(c(c1, c2, c3), nrow = 4)
> contrasts(keyboard$method) <- ctr
>
> keyboard.aov <- aov(score ~ method, data = keyboard)
> summary(keyboard.aov, split = list(method = list("A v B,C,D" = 1, "B v C,D" = 2,
+                                                "C v D,E" = 3)))
```

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
method	3	638.3	212.8	8.343	0.000243 ***
method: A v B,C,D	1	42.0	42.0	1.647	0.207540
method: B v C,D	1	96.3	96.3	3.775	0.059884 .
method: C v D,E	1	500.0	500.0	19.606	8.5e-05 ***
Residuals	36	918.1	25.5		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

According to this approach there is no significant difference between Method A and the others, nor between Method B and Methods C and D. There is a highly significant difference between Methods C and D. (The latter accounts for 78% of the sum of squares for method.)

2. You will need the data frame `tyres` [see solutions to Exercises 6].

```
> c1 <- c(1, 1, -1, -1)
> c2 <- c(1, -1, 1, -1)
> c3 <- c(1, -1, -1, 1)
> ctr <- matrix(c(c1, c2, c3), nrow = 4)
> contrasts(tyres$position) <- ctr
> tyres.aov <- aov(wear ~ position, data = tyres)
> summary(tyres.aov)
```

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
position	3	1189.0	396.3	13.74	6.28e-06 ***
Residuals	32	923.1	28.8		

```
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

> summary(tyres.aov, split = list(position = list("front v rear" = 1, "offside v nearside" = 2,
                                                    "interaction" = 3)))
```

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
position	3	1189.0	396.3	13.739	6.28e-06 ***
position: front v rear	1	1006.8	1006.8	34.901	1.42e-06 ***
position: offside v nearside	1	180.9	180.9	6.272	0.0176 *
position: interaction	1	1.3	1.3	0.045	0.8332
Residuals	32	923.1	28.8		

```
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

There is very strong evidence of overall differences in wear among the four positions for the tyres.

The wear on the rear tyres is significantly greater than on the front tyres, the difference being very highly significant ($p = 0.0000014$). The wear on the offside tyres is significantly greater than on the nearside tyres ($p = 0.0176$). The sums of squares for the corresponding contrasts account for 99.9% of the total sum of squares for position. The third contrast, for interaction, is not significant.