

Introduction to R

Answers to Session 7 exercises

Statistical Consulting Centre

2 March, 2017

1 *t*-tests

Carry out a two-sample *t*-test to determine whether:

- (i) males and females have different mean nerdy scores.

```
with(sports.df, t.test(nerdy.sc~gender))

Welch Two Sample t-test

data:  nerdy.sc by gender
t = -3.8204, df = 965.27, p-value = 0.0001418
alternative hypothesis: true difference in means is not equal to 0
95 percent confidence interval:
 -0.18630715 -0.05985835
sample estimates:
mean in group Female    mean in group Male
      2.949468           3.072551
```

- (ii) the mean nerdy score of respondents living with their partners differs from that of respondents who do not live with their partners.

```
with(sports.df, t.test(nerdy.sc~partner))

Welch Two Sample t-test

data:  nerdy.sc by partner
t = -1.0193, df = 211.86, p-value = 0.3092
alternative hypothesis: true difference in means is not equal to 0
95 percent confidence interval:
 -0.18490690  0.05886231
sample estimates:
mean in group No mean in group Yes
      2.884848           2.947871
```

2 ANOVA

- (i) Perform a one-way anova to test mean nerdy score differs between the three age groups we have been considering.

```
oneway <- with(sports.df, aov(nerdy.sc~age.group))
summary(oneway)
```

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
age.group	2	2.22	1.1081	4.31	0.0137 *
Residuals	986	253.52	0.2571		

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
7 observations deleted due to missingness

- (ii) What are your conclusions from the one-way anova?

At least one age group's mean nerdy score differs from that of the others.

- (iii) Find the estimated mean nerdy score over all age groups and for individual age groups.

```
model.tables(oneway, "means")
```

Tables of means

Grand mean

3.006218

age.group

Under 40 41 to 60 Over 61

3.009 3.059 2.946

rep 293.000 364.000 332.000

- (iv) Perform pair-wise comparisons of mean nerdy scores between all age groups using Tukey's Honest Significance Difference method to compute p -values adjusted for multiple comparisons.

```
TukeyHSD(oneway)
```

Tukey multiple comparisons of means

95% family-wise confidence level

Fit: aov(formula = nerdy.sc ~ age.group)

\$age.group

	diff	lwr	upr	p adj
41 to 60-Under 40	0.04925709	-0.04415998	0.14267415	0.4312230
Over 61-Under 40	-0.06360922	-0.15901291	0.03179448	0.2613921
Over 61-41 to 60	-0.11286630	-0.20319233	-0.02254027	0.0096125

(v) Which pairs of age groups differ in mean nerdy score?

“41–60” and “Over 61”

(vi) Perform a two-way anova of nerdy score on age group and gender.

```
twoway <- with(sports.df, aov(nerdy.sc~age.group*gender))
summary(twoway)
```

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
age.group	2	2.22	1.108	4.384	0.0127	*
gender	1	4.19	4.193	16.588	5.02e-05	***
age.group:gender	2	0.86	0.428	1.693	0.1845	
Residuals	983	248.47	0.253			

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

7 observations deleted due to missingness

(vii) Which rows of the two-way ANOVA table are statistically significant?

Those corresponding to `age.group` and `gender`. The interaction between `age.group` and `gender` is not statistically significant at the 5% level since $\text{Pr}(>F) > 0.05$.

(viii) Calculate the estimated means for each `age.group`, `gender` and `age.group`–`gender` combination. Perform *appropriate* pair-wise comparisons of means.

```
model.tables(twoway, "means")
```

Tables of means

Grand mean

3.006218

age.group

Under 40 41 to 60 Over 61

3.009 3.059 2.946

rep 293.000 364.000 332.000

gender

Female Male

2.946 3.076

rep 533.000 456.000

age.group:gender

gender

age.group Female Male

Under 40 2.99 3.03

rep 171.00 122.00

41 to 60 2.98 3.16

rep 204.00 160.00

Over 61	2.86	3.02
rep	158.00	174.00

TukeyHSD(twoway)

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

Fit: aov(formula = nerdy.sc ~ age.group * gender)

\$age.group

	diff	lwr	upr	p adj
41 to 60-Under 40	0.04925709	-0.04336658	0.14188075	0.4250499
Over 61-Under 40	-0.06360922	-0.15820264	0.03098421	0.2554626
Over 61-41 to 60	-0.11286630	-0.20242518	-0.02330742	0.0088880

\$gender

	diff	lwr	upr	p adj
Male-Female	0.1300681	0.06713236	0.1930038	5.4e-05

\$`age.group:gender`

	diff	lwr	upr	p adj
41 to 60:Female-Under 40:Female	-0.01193670	-0.160777938	0.13690453	
Over 61:Female-Under 40:Female	-0.13137908	-0.289792553	0.02703439	
Under 40:Male-Under 40:Female	0.03953121	-0.130597111	0.20965952	
41 to 60:Male-Under 40:Female	0.16472588	0.006827846	0.32262391	
Over 61:Male-Under 40:Female	0.02933555	-0.125245990	0.18391709	
Over 61:Female-41 to 60:Female	-0.11944238	-0.271578176	0.03269342	
Under 40:Male-41 to 60:Female	0.05146791	-0.112830956	0.21576678	
41 to 60:Male-41 to 60:Female	0.17666258	0.025063566	0.32826160	
Over 61:Male-41 to 60:Female	0.04127226	-0.106869319	0.18941383	
Under 40:Male-Over 61:Female	0.17091029	-0.002107728	0.34392830	
41 to 60:Male-Over 61:Female	0.29610496	0.135097575	0.45711234	
Over 61:Male-Over 61:Female	0.16071463	0.002958359	0.31847091	
41 to 60:Male-Under 40:Male	0.12519467	-0.047351533	0.29774088	
Over 61:Male-Under 40:Male	-0.01019565	-0.179712194	0.15932089	
Over 61:Male-41 to 60:Male	-0.13539033	-0.292629003	0.02184835	
	p adj			
41 to 60:Female-Under 40:Female	0.9999140			
Over 61:Female-Under 40:Female	0.1685811			
Under 40:Male-Under 40:Female	0.9858132			
41 to 60:Male-Under 40:Female	0.0350612			
Over 61:Male-Under 40:Female	0.9944170			
Over 61:Female-41 to 60:Female	0.2196437			
Under 40:Male-41 to 60:Female	0.9478634			
41 to 60:Male-41 to 60:Female	0.0116853			
Over 61:Male-41 to 60:Female	0.9683015			
Under 40:Male-Over 61:Female	0.0550741			
41 to 60:Male-Over 61:Female	0.0000028			
Over 61:Male-Over 61:Female	0.0429638			

41 to 60:Male-Under 40:Male	0.3030179
Over 61:Male-Under 40:Male	0.9999793
Over 61:Male-41 to 60:Male	0.1376475

3 Tests of Independence

- (i) Produce a two-way frequency table of counts between `income` and `gender`.

```
with(sports.df, table(income, gender))
```

income	gender	
	Female	Male
5 000\$	59	27
10 000\$-15 000\$	61	59
15 000\$-20 000\$	57	41
20 000\$-25 000\$	38	38
25 000\$-30 000\$	40	33
30 000\$-40 000\$	67	63
40 000\$-50 000\$	55	51
50 000\$-70 000\$	66	62
70 000\$-100 000\$	31	34
> 100 000\$	30	33

- (ii) Do you think that income level depends on gender? Perform an appropriate test to find out.

```
chisq.test(with(sports.df, table(income, gender)))
```

Pearson's Chi-squared test

```
data: with(sports.df, table(income, gender))
X-squared = 11.756, df = 9, p-value = 0.2274
```