

NZSSN Courses: Introduction to R

Session 7 – Simple analysis

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SCIENCE
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Regression commands

Two of the most commonly used R commands for modeling:

- `lm()`: fits **L**inear **M**odels
- `glm()`: fits **G**eneralised **L**inear **M**odels.

Note SAS users: PROC GLM is **not** the same as R's `glm()`.

There's a lot in these two commands; entire stage 3 statistical courses on linear and generalised linear models.

Student's *t*-test

```
t.test(y ~ x)
```

- *y*: values; e.g., `total.lik`, `Q1.lik`, `Age`, etc.
- *x*: group; e.g., `Gender`, `Q5` (obedient or think themselves).

Suppose we want to test whether males and females (*x* = `Gender`) have different total scores across Q1 – Q4 (*y* = `total.lik`).

Categorical variables should be converted to type `factor` before analysis, i.e.

```
issp.df$Gender <- factor(issp.df$Gender)
with(issp.df, t.test(total.lik~Gender))
```

Student's *t*-test

Welch Two Sample t-test

```
data: total.lik by Gender
```

```
t = 4.3417, df = 874.71, p-value = 1.579e-05
```

```
alternative hypothesis: true difference in means is not equal
```

```
95 percent confidence interval:
```

```
0.3541459 0.9384793
```

```
sample estimates:
```

mean in group Female	mean in group Male
12.71067	12.06436

- The estimated difference in total score between females and males is $12.71 - 12.06 = 0.65$.
- $p\text{-value} = 1.579 \times 10^{-5}$, i.e. we have extremely strong evidence that the mean total score are statistically significantly different.

Student's *t*-test

Welch Two Sample t-test

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data: total.lik by Gender
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```
sample estimates:
```

mean in group Female	mean in group Male
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- While we have statistical significance, we should note that the sample sizes are very large.
- Is the observed difference significant from a social scientist's perspective?

Multiple comparisons

Let's compare the total score between three age groups, i.e.

- 1 Do a t -test between "Under 35" and "36 to 60".
- 2 Do a t -test between "Under 35" and "Over 61".
- 3 Do a t -test between "36 to 60" and "Over 61".

Really?

Error rate

When we do a t -test comparing mean total score between females and males, the null hypothesis is that the mean total score for females is the same as that for males. The t -test is performed (with the hope) to reject this null hypothesis.

In order to come up with a p-value, we *assume* that α (typically 5%) of the time, we will reject the null hypothesis when it's actually true, i.e., we assume 5% of the time we will make a mistake.

- When we do two simultaneous t -tests, about 10% of the time we will make a mistake.
- When we do three simultaneous t -tests, about 15% of the time we will make a mistake.
- The chance of being shot in Russian Roulette is 16.67%. Would you risk it then?

Analysis of Variance (ANOVA)

Generalises *t*-test to more than two groups.

Null hypothesis: all group means are equal.

Example. Mean total score is the same for all three age.groups.

```
tryaov <- with(issp.df, aov(total.lik~age.group))
```

- `aov()`: **A**nalysis of **V**ariance.
- Response variable (i.e. `total.lik`) is separated by `~` from explanatory variable(s) (i.e. `age.group`).
- All explanatory variables should be categorical (otherwise it's not ANOVA).

aov()

```
summary(tryaov)
```

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
age.group	2	803	401.4	89.82	<2e-16 ***
Residuals	951	4250	4.5		

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

93 observations deleted due to missingness

We have extremely strong evidence that at least one age group's mean total score is different to that of the other age groups.

Which one(s) is(are) different????

Which one(s)?

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

Tables of means

Grand mean

12.43711

age.group

	Under 35	36 to 60	Over 61
--	----------	----------	---------

	13.39	12.46	10.79
--	-------	-------	-------

rep	319.00	446.00	189.00
-----	--------	--------	--------

The mean total score...

- over all participants is 12.4.
- for “Under 35” group is higher than both that of the “36 to 60” and the “Over 61” groups.
- for “36 to 60” group is higher than the “Over 61” group.

Which one(s)?

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

Tables of means

Grand mean

12.43711

age.group

	Under 35	36 to 60	Over 61
--	----------	----------	---------

	13.39	12.46	10.79
--	-------	-------	-------

rep	319.00	446.00	189.00
-----	--------	--------	--------

- Are any pairs of these means statistically different from one another?

Post-hoc multiple comparisons

```
TukeyHSD(tryaov)
```

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

```
Fit: aov(formula = total.lik ~ age.group)
```

```
$age.group
```

		diff	lwr	upr	p adj
36 to 60-Under 35	-0.9335578	-1.297433	-0.5696823	0	
Over 61-Under 35	-2.6003549	-3.055857	-2.1448527	0	
Over 61-36 to 60	-1.6667972	-2.097496	-1.2360986	0	

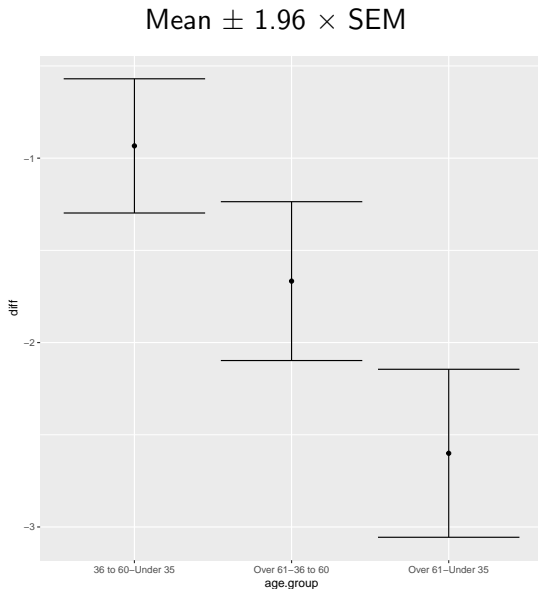
- `diff`: estimated difference between two group means.
- `lwr`, `upr`: lower and upper limit of the 95% confidence interval of the estimated difference.
- `p adj`: p-values adjusted for multiple comparisons.

Post-hoc multiple comparisons

	diff	lwr	upr	p adj
36 to 60-Under 35	-0.9335578	-1.297433	-0.5696823	7.353107e-09
Over 61-Under 35	-2.6003549	-3.055857	-2.1448527	1.399991e-13
Over 61-36 to 60	-1.6667972	-2.097496	-1.2360986	1.690870e-13

- Mean total score for “36 to 60” is 0.9 units (on the likert scale) *lower* than “Under 35” ($p \text{ adj} < 0.0001$).
- Mean total score for “Over 61” is 2.6 units *lower* than “Under 35” ($p \text{ adj} < 0.0001$).
- Mean total score for “Over 61” is 1.7 units *lower* than “36 to 60” ($p \text{ adj} < 0.0001$).

From Session 6: Mean total score vs Age group



Two-way ANOVA

- `tryaov` was fitted using one categorical explanatory variable (`age.group`). We therefore refer to its ANOVA table as *one-way*.
- If we fit a linear model using two categorical explanatory variables, we have a *two-way* ANOVA.
- Recall: All categorical variables should be converted into factors.

```
issp.df$Gender <- factor(issp.df$Gender)
try2way <- with(issp.df,
               aov(total.lik~Gender*age.group))
```

- `Gender*age.group` is equivalent to `Gender + age.group + Gender:age.group`.

Two-way ANOVA

```
summary(try2way)
```

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
Gender	1	98	97.7	22.200	2.82e-06	***
age.group	2	774	386.8	87.905	< 2e-16	***
Gender:age.group	2	15	7.3	1.654	0.192	
Residuals	947	4167	4.4			

```
---
```

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

There is no two-way interaction between Gender and age.group (p -value = 0.19), i.e., the magnitude of the difference in mean total score between males and females is constant across all age groups, and vice versa.

Two-way ANOVA

```
summary(try2way)
```

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
Gender	1	98	97.7	22.200	2.82e-06	***
age.group	2	774	386.8	87.905	< 2e-16	***
Gender:age.group	2	15	7.3	1.654	0.192	
Residuals	947	4167	4.4			

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

94 observations deleted due to missingness

We have extremely strong evidence that:

- the mean total score of *at least one* age group differs from the others, and
- mean total score differs between males and females.

Estimated means

```
model.tables(try2way, "means")
```

Tables of means

Grand mean

12.43757

Gender

	Female	Male
--	--------	------

	12.71	12.06
--	-------	-------

rep	549.00	404.00
-----	--------	--------

age.group

	Under 35	36 to 60	Over 61
--	----------	----------	---------

	13.39	12.43	10.84
--	-------	-------	-------

rep	319.00	445.00	189.00
-----	--------	--------	--------

Post-hoc pairwise comparisons

```
TukeyHSD(try2way)
```

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

```
Fit: aov(formula = total.lik ~ Gender * age.group)
```

```
$Gender
```

	diff	lwr	upr	p adj
Male-Female	-0.6478476	-0.9176817	-0.3780135	2.8e-06

```
$age.group
```

	diff	lwr	upr	p adj
36 to 60-Under 35	-0.9533867	-1.314615	-0.5921584	0
Over 61-Under 35	-2.5488847	-3.000862	-2.0969079	0
Over 61-36 to 60	-1.5954980	-2.023006	-1.1679900	0

Test of independence

```
Q5.age.tab <- with(issp.df, table(Q5, age.group))  
Q5.age.tab
```

	age.group		
Q5	Under 35	36 to 60	Over 61
be obedient	38	74	75
think themselves	259	353	122

Do opinions on preparing children for life depend on age group?
Statistically speaking, is Q5 (the variable) and age.group independent of one another?

Pearson's Chi-squared test

```
chisq.test(Q5.age.tab)
```

Pearson's Chi-squared test

data: Q5.age.tab

X-squared = 51.115, df = 2, p-value = 7.955e-12

- There is extremely strong evidence ($p\text{-value} < 0.0001$) that Q5 and age.group are not independent of one another.
- Opinions on preparing children for life depend on the age group to which respondents belong.

Assumptions

- Pearson's Chi-squared tests have certain assumptions. Beyond the scope of this course.
- `chisq.test()` will give you a warning if these assumptions are not met.

```
Warning in chisq.test(mytest):  Chi-squared  
approximation may be incorrect
```

- These assumptions are more likely to be wrong if the sample size is small.
- If this happens, the alternative is to use Fisher's exact test.

Fisher's exact test

Assume Q5.age.tab does not meet the underlying assumptions of Pearson's Chi-squared test.

```
fisher.test(Q5.age.tab)
```

Fisher's Exact Test for Count Data

```
data: Q5.age.tab  
p-value = 6.93e-11  
alternative hypothesis: two.sided
```

Summary

- Student's t -test
- One-way ANOVA
- Two-way ANOVA
- Pearsons Chi-squared test
- Fishers exact test