

Statistics 151 (Linear Modelling -Theory and Applications)

Homework Two

Due on 02 March, 2015

20 February, 2015

1. Consider simple linear regression where there is one response variable y and one explanatory variable x and there are n subjects with values y_1, \dots, y_n and x_1, \dots, x_n . The model is $y_i = \beta_0 + \beta_1 x_i + e_i$ where e_1, \dots, e_n are independent $N(0, \sigma^2)$. Show that $\hat{\beta}_0$ and $\hat{\beta}_1$ are independent if $x_1 + \dots + x_n = 0$. Here, of course, $\hat{\beta}_0$ and $\hat{\beta}_1$ denote the least squares estimates of β_0 and β_1 .
2. In the Bodyfat dataset, consider the linear model:

$$\text{BODYFAT} = \beta_0 + \beta_1 \text{KNEE} + \beta_2 \text{THIGH} + \beta_3 \text{HIP} + \beta_4 \text{ANKLE} + e$$

Assume that the errors are i.i.d normal.

- (a) Construct an F -test for testing $H_0 : \beta_1 + \beta_2 = \beta_3 + \beta_4$. Describe your method and report the value of the F -statistic, its degrees of freedom and the p -value.
 - (b) Construct a t -test for testing $H_0 : \beta_1 + \beta_2 = \beta_3 + \beta_4$. Describe your method and report the value of the t -statistic, its degrees of freedom and the p -value.
 - (c) How is the value of your t -test statistic related to the value of the F -test statistic?
3. In the following regression output, the value of the F -statistic (last line) and its p -value are missing. Fill them in, providing proper reasoning, based on the available information.

Call:

```
lm(formula = y ~ x1 + x2 + x3 + x4 + x5, data)
```

Residuals:

	Min	1Q	Median	3Q	Max
	-10.9307	-2.8923	-0.3829	3.1778	9.5804

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	1.63014	5.95379	0.274	0.784

```

x1          0.85682    0.05065   16.916   < 2e-16 ***
x2         -2.02587    0.39720   -5.100   6.77e-07 ***
x3          0.04083    0.14899    0.274    0.784
x4         -0.33431    0.08191   -4.082   6.05e-05 ***
x5          0.24481    0.18236    1.342    0.181

```

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

Residual standard error: 4.122 on 246 degrees of freedom

Multiple R-squared: 0.7228, Adjusted R-squared: 0.7172

F-statistic: XXXXX on X and 246 DF, p-value: XXXXX

4. In the linear model, show that the square of the (sample) correlation between the response values (y_1, \dots, y_n) and the fitted values $(\hat{y}_1, \dots, \hat{y}_n)$ equals the coefficient of determination, R^2 .
5. Last year, 80 students took this particular course at Berkeley of whom 20 were freshmen, 20 were sophomores, 20 juniors and 20 seniors. In R, I have saved the scores for the 20 freshmen in the vector `g1`, for the 20 sophomores in `g2`, juniors in `g3` and seniors in `g4`. Consider the following output:

```

> mean(g1)
[1] 58.53768
> sd(g1)
[1] 5.024681
> mean(g2)
[1] 64.72989
> sd(g2)
[1] 4.43851
> mean(g3)
[1] 64.06235
> sd(g3)
[1] 5.264511
> mean(g4)
[1] 66.27922
> sd(g4)
[1] 4.192543

```

The instructor wants to know if these different average scores for the four groups are caused merely by randomness or if there is really a connection between the performance ability of students and their year. Let y_1, \dots, y_n (for $n = 80$) denote the scores of the students. The instructor makes the assumption that these are independent and that y_i is distributed according to $N(\mu_j, \sigma^2)$ if the i th student is in the j th year. She wants to test the hypothesis $H_0 : \mu_1 =$

$\mu_2 = \mu_3 = \mu_4$ against its complement H_1 . Following the steps outlined below, show that this test can be carried out via the F-test that we learned for the linear model.

- Define four explanatory variables x_1, x_2, x_3 and x_4 in the following way: x_j takes the value $x_{ij} = 1$ for the i th subject if the i th subject is in year j ; otherwise x_j takes the value $x_{ij} = 0$. Show that $y_i \sim N(\mu_i, \sigma^2)$ is equivalent to the statement that $y_i = \mu_1 x_{i1} + \mu_2 x_{i2} + \mu_3 x_{i3} + \mu_4 x_{i4} + e_i$.
- Calculate the RSS in this linear model.
- Calculate the RSS in the reduced model under the constraint $\mu_1 = \mu_2 = \mu_3 = \mu_4$.
- Calculate the p -value for the F-test.
- Is there enough evidence in this data to reject the instructor's null hypothesis?

6. Consider the following R output:

Call:

```
lm(formula = y ~ x1 + x2 + x3 + x4 + x5, data)
```

Residuals:

Min	1Q	Median	3Q	Max
-17.3214	-3.8831	-0.0002	3.6401	16.1967

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	-0.74860	10.34432	-0.072	0.9424
x1	0.18853	0.03039	XXXXX	XXXX
x2	-15.17748	32.12529	XXXXX	XXXX
x3	15.30167	32.12624	XXXXX	XXXX
x4	-0.45922	0.10500	-4.374	1.81e-05 ***
x5	0.35741	0.15070	2.372	0.0185 *

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

Residual standard error: 5.337 on 246 degrees of freedom

Multiple R-squared: 0.5353, Adjusted R-squared: 0.5259

F-statistic: 56.68 on 5 and 246 DF, p-value: < 2.2e-16

- What is the p -value for the F-test for testing $H_0 : \beta_2 = 0$?
- What is the p -value for the F-test for testing $H_0 : \beta_3 = 0$?
- What is the p -value for the F-test for testing $H_0 : \beta_2 = \beta_3 = 0$? You may use information from the following R output corresponding to the same dataset as above.

Call:

```
lm(formula = y ~ x1 + x4 + x5, data)
```

Residuals:

	Min	1Q	Median	3Q	Max
	-12.8578	-4.0721	-0.0354	3.6837	20.1068

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	-32.38716	7.93171	-4.083	6.00e-05 ***
x1	0.24649	0.02860	8.619	8.09e-16 ***
x4	-0.24999	0.09748	-2.564	0.0109 *
x5	0.97294	0.06839	14.227	< 2e-16 ***

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

Residual standard error: 5.534 on 248 degrees of freedom

Multiple R-squared: 0.4963, Adjusted R-squared: 0.4902

F-statistic: 81.46 on 3 and 248 DF, p-value: < 2.2e-16