

## MATH371301

This question paper consists of 7 printed pages, each of which is identified by the reference **MATH3713**.

Graph paper is provided. Only approved basic scientific calculators may be used.

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Examination for the Module MATH3713  
(January 2006)

## Regression and Smoothing

Time allowed: **3 hours**

Attempt not more than FOUR questions.  
All questions carry equal marks.

1. The multiple linear regression model involving  $n$  observations and  $k$  explanatory variables can be written in the form

$$y_i = \beta_0 + \sum \beta_j x_{ij} + \epsilon_i.$$

- (a) Explain each of the terms in this model, the assumptions on the  $\{\epsilon_i\}$  and the ranges for the indices  $i$  and  $j$ .
- (b) The maximum likelihood estimator of the regression parameter  $\beta$  can be written in matrix form as  $\hat{\beta} = CX^T\mathbf{y}$ . Verify this formula, taking care to define and give the dimensions of  $C$ ,  $X$  and  $\mathbf{y}$ .
- (c) A scientist wishes to explore the dependence of January rainfall (in mm) on mean January temperature in Leeds (in °C) by fitting a simple linear regression ( $k = 1$ ) using 20 years of measurements. He obtains estimated regression coefficients  $\hat{\beta}_0 = 10.2$  and  $\hat{\beta}_1 = 2.4$ , with estimated standard deviations 2.2 and 1.3, respectively.
- (i) Find the associated  $t$ -statistic for  $\hat{\beta}_1$ . Describe what hypothesis can be tested with this  $t$ -statistic and state what conclusions can be reached in this example.
- (ii) A second scientist repeats the analysis on the same dataset but in different units, using inches for rainfall and °F for temperature. Give the new values of the estimated regression coefficients, and give the new standard error and  $t$ -statistic for  $\hat{\beta}_1$ . What effect does this change in units have on the conclusion reached in part (i)?

[Hints: (1) Note that 1 inch equals 25.4 mm. (2) If  $C$  denotes a temperature on the Celsius scale (°C), then the corresponding value  $F$  on the Fahrenheit scale (°F) is given by  $F = 32 + (9/5)C$ . (3) The following R output gives some upper 2.5% critical values for the  $t$ -distribution.]

```
round(qt(.975,c(15,16,17,18,19,20)),3)
[1] 2.131 2.120 2.110 2.101 2.093 2.086
```

2. The following R output gives the result of fitting a multiple linear regression model. Here  $y$ ,  $x_1$ ,  $x_2$  are all numeric vectors of the same length.

```
> lm1=lm(y~x1+x2)
> summary(lm1)
```

Call:

```
lm(formula = y ~ x1 + x2)
```

Residuals:

Min	1Q	Median	3Q	Max
-1.8942	-0.5667	0.1963	0.7334	1.7128

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	-0.2684	0.9637	-0.279	0.787
x1	3.2793	0.3328	A	4.04e-06 ***
x2	1.8909	0.2319	8.153	1.90e-05 ***

---

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

Residual standard error: 1.263 on 9 degrees of freedom

Multiple R-Squared: 0.9913, Adjusted R-squared: B

F-statistic: 511.6 on 2 and 9 DF, p-value: 5.395e-10

```
> anova(lm1)
```

Analysis of Variance Table

Response: y

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
x1	1	1525.50	1525.50	956.803	1.890e-10 ***
x2	1	105.98	105.98	C	1.902e-05 ***
Residuals	D	14.35	1.59		

---

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

>

- (a) In this R output, 4 entries labelled A – D are missing. Give these values. Also give the sample size of the dataset and the values of  $SS_T$ ,  $SS_R$ ,  $SS_E$ , the “total”, “regression” and “error” sums of squares, respectively.
- (b) For the multiple linear regression model, give the definitions for the three types of residuals  $e_i$ ,  $r_i$ , and  $t_i$  (raw, standardized and studentized, respectively).
- (c) The statistic  $DFFITS_i$  is defined by

$$DFFITS_i = (\hat{y}_i - \hat{y}_{i,-i}) / (\hat{\sigma}_{-i} \sqrt{h_{ii}}).$$

Explain this notation and motivate the definition.

It is known that  $\text{DFFITs}_i$  can also be written in the form

$$\text{DFFITs}_i = \left( \frac{h_{ii}}{1 - h_{ii}} \right)^{1/2} t_i.$$

Using the second representation, explain how this statistic combines the effects of leverage and outlyingness.

If  $h_{11} = 0.2$  and  $t_1 = 6$  for the dataset in (a), what conclusions would you draw?

3. Consider a multiple linear regression model with  $k > 1$  regressor variables  $x_1, \dots, x_k$  on  $n$  observations.

- (a) Why is variable selection an important problem? Explain the method of forward search, describing, in particular, how new variables are added during the algorithm.
- (b) Consider a hypothesis test to compare the models

$$M_0 : y \sim x_1 \text{ vs. } M_1 : y \sim x_1 + x_2.$$

Using the error sum of squares for the two models (labelled  $SS_E(1)$  and  $SS_E(1, 2)$ , say), show how an  $F$  statistic with 1 and  $n-3$  degrees of freedom can be constructed to carry out the test.

- (c) In a dataset involving a response variable  $y$  and three possible regressor variables  $x_1, x_2, x_3$  on  $n = 25$  observations, the following table lists the error sums of squares ( $SS_E$ ) for a collection of models. The entry under “Model” lists the regressor variables in the model. For the top row, only an intercept is present.

Model	$SS_E$
-	88
1	52
2	48
3	82
12	46
13	29
23	24
123	18

Carry out one step of forwards search and backwards elimination on this dataset, and compare the models chosen in each case.

[Hint: The following R output gives some upper 5% critical values for the F distribution.]

```
> round(qf(.95, 1, c(20, 21, 22, 23, 24, 25)), 2)
[1] 4.35 4.32 4.30 4.28 4.26 4.24
```

4. (a) Consider the family of transformations indexed by a real parameter  $\lambda$ , defined by

$$f_\lambda(x) = \frac{x^\lambda - 1}{\lambda}, \quad x > 0, \quad \lambda \neq 0$$

and by

$$f_0(x) = \log(x), \quad x > 0.$$

Show that for  $\lambda < 1$ ,  $f_\lambda(x)$  is increasing and concave in  $x > 0$  with  $f'(1) = 1$ . What happens if  $\lambda > 1$ ?

- (b) Suppose that in a regression analysis of a real-valued response variable  $y$  on a real-valued explanatory variable  $x$ , a plot of  $y$  vs.  $x$  suggests that  $y$  is approximately a decreasing convex function of  $x$ . Discuss (i) conditions under which applying the transformation in (a) to  $y$  or  $x$  might be useful, (ii) what values of  $\lambda$  might be appropriate, (iii) why such a transformation might be useful, and (iv) how you might assess the suitability of a particular transformation.
- (c) In a study to investigate the effectiveness of slug pellets, various dosages ( $x$  = number of pellets per square metre) were sprinkled on a selection of identically-sized English country gardens one evening, and the numbers ( $y$ ) of dead slugs the next morning were counted. A plot of  $y$  vs.  $x$  is given in Figure 1. There were three replications at each of 4 doses, so  $n = 12$ .

A statistician considers three models  $M_1 : y \sim x$ ,  $M_2 : y \sim x^{1/3}$  and  $M_3 : y \sim x + x^{1/3}$ . On the basis of the information in the following *R* output, investigate which of the models seems to be most appropriate.

```
> x
[1] 25 25 25 50 50 50 100 100 100 200 200 200
> y
[1] 332 295 330 337 284 395 545 457 426 560 464 564
> lm1=lm(y~x)
> summary(lm1)
```

```
Call:
lm(formula = y ~ x)
```

```
Residuals:
    Min       1Q   Median       3Q      Max
-84.609 -26.728   2.326  19.902 121.435
```

```
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept) 298.5217    28.1999   10.59 9.41e-07 ***
x              1.2504     0.2447    5.11 0.000457 ***
---

```

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

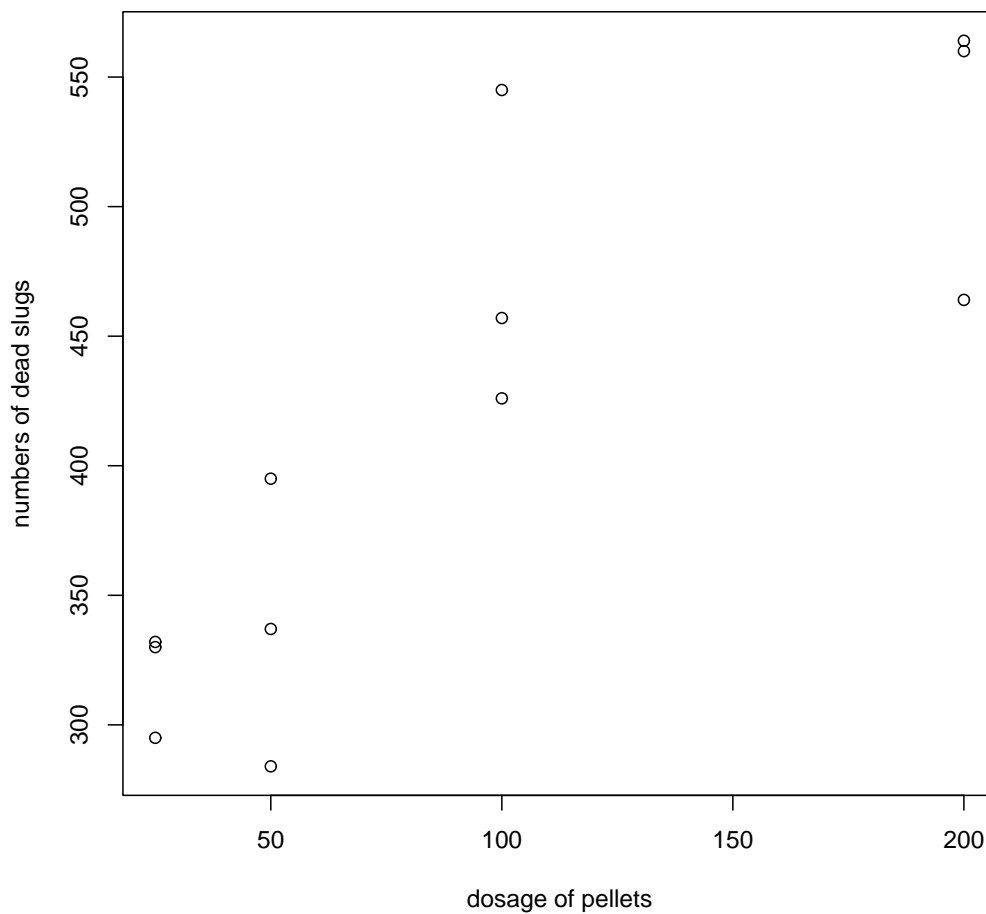


Figure 1: Plot for Question 4: numbers of dead slugs vs. dosage of slug pellets.

Residual standard error: 56.81 on 10 degrees of freedom  
 Multiple R-Squared: 0.7231, Adjusted R-squared: 0.6954  
 F-statistic: 26.11 on 1 and 10 DF, p-value: 0.0004572

```
> x3=x^(1/3)
> lm2=lm(y~x3)
> summary(lm2)
```

```
Call:
lm(formula = y ~ x3)
```

```
Residuals:
    Min     1Q  Median     3Q     Max
-85.22 -22.07  16.26  23.07 100.31
```

```

Coefficients:
              Estimate Std. Error t value Pr(>|t|)
(Intercept)    78.84      60.98   1.293 0.225160
x3              78.82      13.82   5.703 0.000198 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 52.36 on 10 degrees of freedom
Multiple R-Squared:  0.7648,    Adjusted R-squared:  0.7413
F-statistic: 32.52 on 1 and 10 DF,  p-value: 0.0001976

> lm3=lm(y~x+x3)
> summary(lm3)

Call:
lm(formula = y ~ x + x3)

Residuals:
    Min       1Q   Median       3Q      Max
-88.50 -27.67  14.22  26.25  93.94

Coefficients:
              Estimate Std. Error t value Pr(>|t|)
(Intercept)  13.0818    219.8512   0.060   0.954
x             -0.4015     1.2845  -0.313   0.762
x3            103.0109    78.7292   1.308   0.223

Residual standard error: 54.89 on 9 degrees of freedom
Multiple R-Squared:  0.7674,    Adjusted R-squared:  0.7157
F-statistic: 14.84 on 2 and 9 DF,  p-value: 0.001413

```

5. Given data points  $x_1 < \dots < x_n$  and a bandwidth parameter  $h > 0$ , the kernel density estimate (KDE) is defined by the function

$$\hat{f}_h(x) = \frac{1}{nh} \sum_{i=1}^n K((x - x_i)/h).$$

- What regularity conditions are usually assumed for the kernel  $K(\cdot)$  and what probability model is usually assumed for the data? Make these assumptions for the remainder of the question.
- Find the limit of the  $\hat{f}_h(x)$  as  $n \rightarrow \infty$  for fixed  $h$  and  $x$ .
- For fixed data and assuming  $x = x_i$  for some  $i = 1, \dots, n$ , so that  $x$  equals one of

the data points, show that

$$\hat{f}_h(x) \rightarrow \infty \text{ as } h \rightarrow 0, \text{ and}$$

$$\hat{f}_h(x) \rightarrow 0 \text{ as } h \rightarrow \infty.$$

What is the significance of these results for data analysis?

- (d) In the asymptotic theory of kernel density estimation, under the conditions  $n \rightarrow \infty$ ,  $h \rightarrow 0$  and  $nh \rightarrow \infty$ , it can be shown that the bias and variance of  $\hat{f}_h(x)$  for fixed  $x$  can be approximated by

$$\frac{h^2}{2}A(x) \text{ and } (nh)^{-1}B(x)$$

for certain functions  $A(x)$  and  $B(x)$ , where  $B(x) > 0$ . Using these results describe the asymptotic behaviour of the mean squared error of  $\hat{f}_h(x)$  for fixed  $x$  under the following settings:

- (i)  $h = n^{-4/5}$ ,
- (ii)  $h = n^{-1/5}$ ,
- (iii)  $h = n^{-1/10}$ .

Assuming  $A(x) \neq 0$ , show that choice (ii) is better than the others. Find a constant  $c$  (depending on  $x$ ) such that  $h = ch^{-1/5}$  gives an improvement on choice (ii).

END