

# M. PHIL. IN STATISTICAL SCIENCE

Tuesday 6 June 2006 1.30 to 4.30

# APPLIED STATISTICS

Attempt **FOUR** questions. There are **FIVE** questions in total. The questions carry equal weight.

**STATIONERY REQUIREMENTS** Cover sheet Treasury Tag Script paper **SPECIAL REQUIREMENTS** None

You may not start to read the questions printed on the subsequent pages until instructed to do so by the Invigilator.

1 Suppose  $\mathbf{Y} = X\boldsymbol{\beta} + \boldsymbol{\epsilon}$  where  $\mathbf{Y}^T = (Y_1, \dots, Y_n), \ \boldsymbol{\beta}^T = (\beta_1, \dots, \beta_p), X$  is known  $n \times p$  matrix with rank  $p(\langle n)$ , and  $\boldsymbol{\epsilon}^T = (\epsilon_1, \dots, \epsilon_n)$ , where  $\epsilon_1, \dots, \epsilon_n$  are independent normal random variables with mean 0 and variance  $\sigma^2$ . Obtain the least squares estimator  $\hat{\boldsymbol{\beta}}$  of  $\boldsymbol{\beta}$ , and derive the distribution of  $\hat{\boldsymbol{\beta}}$ .

Let  $\hat{\boldsymbol{\epsilon}} = \mathbf{Y} - \hat{\mathbf{Y}}$  be the vector of residuals, where the vector of fitted values is  $\hat{\mathbf{Y}} = H\mathbf{Y}$  for a matrix H that you should identify. Show that the residuals are uncorrelated with  $\hat{\mathbf{Y}}$ , and uncorrelated with  $\hat{\boldsymbol{\beta}}$ .

[You may assume that for a  $n \times 1$  random vector **Y** and  $n \times n$  matrices A and B,  $\operatorname{cov}(A\mathbf{Y}, B\mathbf{Y}) = A \operatorname{cov}(\mathbf{Y})B^T$ ].

For i = 1, ..., n, let  $Y_i$  be the yield in litres of a chemical experiment and suppose

$$Y_i = \alpha + \beta x_i + \gamma x_i^2 + \epsilon_i ,$$

where  $x_i = (t_i - 100)/10$  and  $t_i$  is the temperature in degrees celsius used in the  $i^{\text{th}}$  experiment. Suppose n = 30 and that  $x_1 = \ldots = x_{10} = -1$ ,  $x_{11} = \ldots = x_{20} = 0$ ,  $x_{21} = \ldots = x_{30} = 1$ . Obtain the covariance matrix of the least squares estimator  $(\hat{\alpha}, \hat{\beta}, \hat{\gamma})^T$ , and find an expression for the variance of  $\hat{\alpha} + \hat{\beta}x + \hat{\gamma}x^2$ . Find the values of x in [-1, 1] for which this variance is (i) maximised and (ii) minimised.

Given that the residual sum of squares is 2.43, derive a 95% confidence interval for the expected yield when the temperature is  $100^{\circ}$ C.

[You may assume that for  $b \neq 0$ ,  $a \neq b$ 

$$\begin{bmatrix} a & 0 & b \\ 0 & b & 0 \\ b & 0 & b \end{bmatrix}^{-1} = \frac{1}{b(a-b)} \begin{bmatrix} b & 0 & -b \\ 0 & a-b & 0 \\ -b & 0 & a \end{bmatrix}.$$

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**2** Explain what is meant by a *factor* in analysis of variance. Suppose a response variable depends on two factors A and B. Define the interaction between A and B.

The edited S-Plus output below shows part of an analysis of the results of an experiment investigating how three factors affect the lustre value of plastic film. The three factors are the thickness of the film (with levels 0 and 1 corresponding to thin and thick), the temperature (0 and 1 corresponding to low and high) of the wash used in preparation of the film, and the length of the wash (1, 2, 3, 4 corresponding to 20, 30, 40, 60 minutes). Assume that the S-Plus objects Thickness, Temperature and Wash have been correctly set up as these factors.

Give the algebraic form and assumptions for the model film1.lm. State the missing degrees of freedom in the output to the directive anova(film1.lm), and explain carefully why the model film2.lm is fitted next.

Find the fitted values under this model and explain how to find the standard errors of these fitted values. Summarise briefly, with appropriate sketch graph(s) as required, how lustre depends on the three factors.

```
> lustre
 [1] 3.3 4.1 4.9 5.0 3.4 4.0
                                  4.2 4.9 19.6 17.5 17.6 20.9 14.5 17.0 15.2
[16] 17.1 5.5
                                  6.1 5.7 6.0 26.6 31.6 30.5 31.4 29.5 30.2
               5.7 5.6 7.2 3.7
[31] 30.2 29.6
> Thickness
 > Temperature
 [1] 0000000111111110000000011111111
> Wash
 [1] 1 2 3 4 1 2 3 4 1 2 3 4 1 2 3 4 1 2 3 4 1 2 3 4 1 2 3 4 1 2 3 4 1 2 3 4
> options(contrasts=c("contr.treatment","contr.poly"))
> film1.lm <- lm(lustre~Thickness*Temperature*Wash)</pre>
> anova(film1.lm)
Analysis of Variance Table
Response: lustre
Terms added sequentially (first to last)
                         Df Sum of Sq
                                       Mean Sq
                                               F Value
                                                           Pr(F)
                Thickness
                              391.300
                                       391.300
                                               191.082 0.0000000
                             2806.878 2806.878 1370.671 0.0000000
              Temperature
                          *
                    Wash
                          *
                               16.443
                                        5.481
                                                 2.677 0.0821499
                                               119.522 0.0000000
                              244.758
                                       244.758
    Thickness:Temperature
                          *
           Thickness:Wash
                                3.568
                                         1.189
                                                 0.581 0.6360491
         Temperature:Wash
                                        0.285
                          *
                                0.856
                                                 0.139 0.9349992
Thickness:Temperature:Wash
                                3.466
                                         1.155
                                                 0.564 0.6464399
                          *
                Residuals
                          *
                               32.765
                                         2.048
> film2.lm <- lm(lustre~Thickness*Temperature)</pre>
> summary(film2.lm,cor=F)
Call: lm(formula = lustre ~ Thickness * Temperature)
Residuals:
  Min
           1Q Median
                        30
                             Max
 -3.35 -0.3687 0.0125 0.5813 3.475
Coefficients:
                      Value Std. Error t value Pr(>|t|)
         (Intercept)
                     4.2250 0.5049
                                        8.3683 0.0000
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Thickness	1.4625	0.7140	2.0483	0.0500
Temperature	13.2000	0.7140	18.4871	0.0000
Thickness:Temperature	11.0625	1.0098	10.9555	0.0000

Residual standard error: 1.428 on 28 degrees of freedom Multiple R-Squared: 0.9837 F-statistic: 562.8 on 3 and 28 degrees of freedom, the p-value is 0

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**3** The random variable Y has a Poisson distribution with mean  $\mu$ . Show that  $\mathbb{P}(Y = y)$  can be written in the form

$$\exp\left\{\frac{y\theta-b(\theta)}{\phi}+c(y,\phi)\right\},$$

and identify  $\theta, b(\theta)$  and  $\phi$ . Verify that  $b'(\theta) = \mathbb{E}(Y)$  and  $\phi b''(\theta) = \operatorname{var}(Y)$ . Explain what is meant by a generalised linear model for the distribution of Y. What is the canonical link function for the Poisson distribution?

Car insurance claims in a particular year are classified according to the class and merit rating of the policyholder. The four merit ratings are

- 3 no claims for 3 or more years
- 2 no claims for 2 years
- 1 no claims for 1 year
- other.

There are five classes

- non-business, no male driver under 25
- 2 non-business, secondary (but not principal) driver is male, under 25
- 3 business
- 4 non-business, principal driver is male, under 25 and unmarried
- 5 non-business, principal driver is male, under 25 and married

Let  $Y_{ij}$  and  $n_{ij}$  be the number of claims and the number of car years insured in class i with merit rating j, for i = 1, ..., 5, j = 0, 1, 2, 3. Consider the rate  $\lambda_{ij}$  of claims per car year insured in class i with merit rating j.

Write down the algebraic form, together with the assumptions, of the model carins.glm in the (slightly edited) S-Plus output below where claims and insured are the number of claims and the claim years insured respectively. Derive equations satisfied by the maximum likelihood estimates of the parameters in the model. Interpret carefully the rest of the S-Plus output.

```
> car <- read.table("carinsurance",header=T)</pre>
>
 car
   merit class insured claims
               1 2757520 217151
 1
        З
 2
        З
               2
                  130535
                           14506
 3
4
        3
               3
                  247424
                           31964
        3
               4
                  156871
                            22884
 5
        3
2
               5
                   64130
                             6560
 6
               1
                  130706
                            13792
 7
        2
               2
                     7233
                             1001
        2
 8
               3
                   15868
                             2695
        2
9
               4
                   17707
                             3054
10
        2
               5
                     4039
                              487
                            19346
11
        1
               1
                  163544
               2
                     9726
                             1430
12
        1
               3
13
        1
                   20369
                             3546
```

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#### **[TURN OVER**

14 1 4 21089 3618 15 5 4869 613 1 16 0 1 273944 37730 0 17 2 21504 3421 3 18 0 37666 7565 19 0 4 56730 11345 0 20 5 8601 1291 > attach(car) > merit <- factor(merit)
> class <- factor(class)</pre> > options(contrasts=c("contr.treatment","contr.poly")) > carins.glm <- glm(claims~offset(log(insured))+merit+class,poisson)</pre> > summary(carins.glm,cor=F) Deviance Residuals: 1Q ЗQ Min Median Max -10.79274 -3.007873 -1.575749 2.426679 11.62523 Coefficients: Value Std. Error t value (Intercept) -2.0357359 0.004311305 -472.18556 merit1 -0.1377590 0.007172219 -19.20730 merit2 -0.2206796 0.007997189 -27.59465 merit3 -0.4929506 0.004502371 -109.48689 class2 0.2998302 0.007258049 41.31003 class3 0.4690550 0.005039141 93.08233 class4 0.5258551 0.005364533 98.02439 class5 0.2155504 0.010734511 20.08013 (Dispersion Parameter for Poisson family taken to be 1 ) Null Deviance: 33854.16 on 19 degrees of freedom Residual Deviance: 579.5163 on 12 degrees of freedom Number of Fisher Scoring Iterations: 3

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4 Let  $Y_1, \ldots, Y_m$  be independent random variables with  $Y_i \sim \text{Bin}(n_i, p_i)$ ,  $i = 1, \ldots, m$ , and consider the model  $\omega_1 : \log\left(\frac{p_i}{1-p_i}\right) = \mathbf{x}_i^T \boldsymbol{\beta}$ ,  $i = 1, \ldots, m$ , where  $\boldsymbol{\beta} = (\beta_1, \ldots, \beta_p)^T$  and  $\mathbf{x}_i$  is a vector of covariate values for  $Y_i$ . Derive an expression for the deviance  $D_1$  for this model. Find an expression for the deviance  $D_0$  for the model  $\omega_0 : \log\left(\frac{p_i}{1-p_i}\right) = \tilde{\mathbf{x}}_i^T \tilde{\boldsymbol{\beta}}$ ,  $i = 1, \ldots, m$ , where  $\tilde{\boldsymbol{\beta}} = (\beta_1, \ldots, \beta_k)^T$  for some fixed  $k, 1 \leq k < p$ , and where  $\tilde{\mathbf{x}}_i$  is the vector of corresponding covariate values. Explain how  $D_0$  and  $D_1$  may be used to test  $H_0 : \beta_{k+1} = \ldots = \beta_p = 0$ .

Anthers of a particular plant species are prepared under two different storage conditions (1, 2 are control, treatment respectively) and with different centrifuging forces (40g, 150g, 350g), and the number of embryogenic anthers is observed. Suppose  $n_{ij}$  anthers are prepared with storage condition i and centrifuging force  $x_j$ ,  $i = 1, 2, x_1 = 40, x_2 = 150, x_3 = 350$ , where  $n_{ij} \ge 70$  for all i, j. Let  $y_{ij}$  and  $p_{ij}$  be the number of embryogenic anthers and probability of the anthers being embryogenic with storage condition i in centrifuging force  $x_j$ , for i = 1, 2, j = 1, 2, 3.

The following models are fitted, resulting in the residual deviances as shown below. In each case i = 1, 2 and j = 1, 2, 3.

$$\omega_{0} : \log\left(\frac{p_{ij}}{1-p_{ij}}\right) = \alpha, \qquad \text{residual deviance } D_{0} = 10.452$$
  

$$\omega_{1} : \log\left(\frac{p_{ij}}{1-p_{ij}}\right) = \alpha_{i}, \qquad \text{residual deviance } D_{1} = 5.173$$
  

$$\omega_{2} : \log\left(\frac{p_{ij}}{1-p_{ij}}\right) = \alpha + \beta \log x_{j}, \qquad \text{residual deviance } D_{2} = 8.092$$
  

$$\omega_{3} : \log\left(\frac{p_{ij}}{1-p_{ij}}\right) = \alpha_{i} + \beta \log x_{j}, \qquad \text{residual deviance } D_{3} = 2.619$$

Interpret these four models. Carry out appropriate statistical tests to determine which is your preferred model.

[If  $\chi^2_{\nu}(\alpha)$  is defined by  $\mathbb{P}(Y > \chi^2_{\nu}(\alpha)) = \alpha$  where  $Y \sim \chi^2_{\nu}$ , then  $\chi^2_1(0.05) = 3.84$ ,  $\chi^2_2(0.05) = 5.99$ ,  $\chi^2_3(0.05) = 7.82$ ,  $\chi^2_4(0.05) = 9.49$ ,  $\chi^2_5(0.05) = 11.07$ ]

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**5** Describe briefly the basic concept behind the E-M algorithm, explaining what the E and M steps correspond to in the algorithm and appropriately defining all notations used.

As part of a training session, a swimming instructor matches into 8 pairs a strong swimmer and a not so strong swimmer. The instructor asks the 8 swimming pairs to each line up behind one of the eight start platforms of a twenty five-metre pool. The instructor then tells the swimmers that there will be a fifty-metre relay race between the pairs. The order that the members of the pairs will race in is left up to each pair to decide. The instructor selects eight further members of his swimming class to record the swimming times for each member of each pair. The race is then started. Unfortunately, five of the eight members asked to record the swimming times of the swimmers only record the total times for their five designated pairs.

The swimming instructor comes to you with the data in the following form

$$X = (S_{11}, S_{12}, S_{21}, S_{22}, S_{31}, S_{32}, T_4, T_5, T_6, T_7, T_8)$$

where  $S_{i1}, S_{i2}$  correspond respectively to the swimming times of the strong swimmer and the not so strong swimmer in the *i*th pair, for  $1 \leq i \leq 3$ , and  $T_j$  corresponds to the total swim time for the *j*th pair, for  $4 \leq j \leq 8$ .

The instructor believes that all swimming times are independent, that those for the strong swimmers follow an exponential distribution with mean  $1/\lambda_1$ , and that those for the weaker swimmers follow an exponential distribution with mean  $1/\lambda_2$ , where  $\lambda_1 > \lambda_2$ . The instructor asks you to use X to estimate  $(\lambda_1, \lambda_2)$ .

(a) Show that for  $j = 4, ..., 8, T_j$  has probability density function

$$g(t) = \frac{\lambda_1 \lambda_2}{\lambda_1 - \lambda_2} \left( e^{-\lambda_2 t} - e^{-\lambda_1 t} \right) \quad \text{for} \quad t > 0.$$

(b) Write down  $\log L(\lambda_1, \lambda_2 | X)$ , the log-likelihood corresponding to the observed data X.

(c) Apply the E-M algorithm to this estimation problem, providing explicit expressions for the parameter updates obtained in the *M*-step of the algorithm to compute the maximum likelihood estimates of  $\lambda_1, \lambda_2$ .

[Note that

$$\int_0^t s \, e^{-\lambda s} \, ds = \frac{1}{\lambda^2} - \left(\frac{t}{\lambda} + \frac{1}{\lambda^2}\right) e^{-\lambda t}$$

for any t > 0 and any  $\lambda$ .]

# END OF PAPER

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