# MAS473 Extended linear models 2013-14 Exam Solutions

(1)(i) Model is

$$Y_{ijk} = \beta_i + b_j + b_{ij} + \epsilon_{ijk},$$

#### [1 mark]

where

 $Y_{ijk}$  is the blood clotting time for drug i, volunteer j and replicate k, i = 1, ..., 3, j = 1, ..., 10 and k = 1, 2

#### [1 mark]

 $b_j \sim N(0, \sigma_1^2)$ 

 $b_{ij} \sim N(0, \sigma_2^2)$ 

 $\epsilon_{ijk} \sim N(0, \sigma^2).$ 

[1 mark]

(1)(ii)

$$\hat{\sigma}_1^2 = 0.001150,$$

 $\hat{\sigma}_2^2 = 0.030827,$ 

 $\hat{\sigma}^2 = 0.008216.$ 

[1 mark]

(1)(iii)

$$\widehat{Var}(Y_{ijk}) = \widehat{Var}(b_i) + \widehat{Var}(b_{ij}) + \widehat{Var}(\epsilon_{ijk}) = 0.040193.$$

[1 mark]

(1)(iv)

$$\widehat{Cov}(Y_{ijk}, Y_{ijk'}) = \widehat{Cov}(b_i + b_{ij} + \epsilon_{ijk}, b_i + b_{ij} + \epsilon_{ijk'}) 
= \widehat{Var}(b_i) + \widehat{Var}(b_{ij}), 
\Rightarrow \widehat{Cor}(Y_{ijk}, Y_{ijk'}) = (0.001150 + 0.030827)/0.040193 = 0.796.$$

[M1 A1]

(1)(v)(a)

$$Var(\bar{Y}_{i..}) = Var\left(\frac{1}{20}\sum_{j=1}^{10}\sum_{k=1}^{2}(\beta_{i} + b_{j} + b_{ij} + \epsilon_{ijk})\right)$$
$$= \frac{Var(b_{i})}{10} + \frac{Var(b_{ij})}{10} + \frac{Var(\epsilon_{ijk})}{20}$$

[2 marks]

Estimated standard error is

$$\sqrt{\frac{0.001150}{10} + \frac{0.030827}{10} + \frac{0.008216}{20}} = 0.06007$$

[1 mark]

(1)(v)(b)

$$Cov(\bar{Y}_{i..}, \bar{Y}_{i'..}) = \frac{1}{400}Cov\left(\sum_{j=1}^{10}\sum_{k=1}^{2}(\beta_i + b_j + b_{ij} + \epsilon_{ijk}), \sum_{j=1}^{10}\sum_{k=1}^{2}(\beta_{i'} + b_j + b_{i'j} + \epsilon_{i'jk})\right)$$

$$= \frac{1}{400}Cov\left(\sum_{j=1}^{10}\sum_{k=1}^{2}b_j, \sum_{j=1}^{10}\sum_{k=1}^{2}b_j\right)$$

$$= \frac{Var(b_j)}{10}$$

[3 marks]

Estimated correlation is

$$\frac{0.001150/10}{0.06007^2} = 0.032.$$

[1 mark]

(1)(v)(c) The standard error is smaller, because under a fixed effect model, there is no variability in the mean of the drug 1 observations due to volunteer or volunteer-drug interaction random effects. [1 mark]

In the fixed effects model, the drug1 term is the expected clotting time, averaged over the 10 volunteers in the study.

[1 mark]

In the mixed effects model, the drug1 term is the expected clotting time, averaged over the population of volunteers from which the 10 were drawn.

[1 mark]

(1)(vi) Test hypothesis  $H_0: \sigma_{b_2}^2 = 0$ .

### [1 mark]

Using GLRT, we compute

$$-2(9.289661 - 22.15275) = 24.189,$$

greater than  $\chi^2_{1;0.99}$ , so strong evidence against  $H_0$ . Conclude that there is evidence of significant random interaction effects between drug and volunteer: the difference in blood clotting times between drugs will vary between volunteers. [1 mark for test, 1 for conclusion]

- (2)(i)  $E(y_i) = exp(\eta_i)/(1 + exp(\eta_i))$ [1 mark]  $\eta_i = \beta_0 + \beta_1 s_i + \beta_2 t_i + \beta_3 s_i t_i$ [1 mark] where  $s_i$  is 1 if the ith surface is B (0 otherwise) and  $t_i$  is the thickness for the ith tyre. [1 mark].
- (2)(ii)  $n_1, ..., n_5 = 100, 50, 40, 50, 50$  [1 mark]  $y_1, ..., y_5 = 75/100, 25/50, 11/40, 10/50, 3/50$  [1 mark]
- (2)(iiia) With  $\eta_i = \beta_0 + \beta_1 s_i$ : Test 1,  $H_0: \beta_1 = 0$   $H_1: \beta_1 \neq 0$ .  $\Delta D = 5.82 > 3.84$  so  $H_0$  rejected. [1 mark] With  $\eta_i = \beta_0 + \beta_1 s_i + \beta_2 t_i$ : Test 2,  $H_0: \beta_2 = 0$   $H_1: \beta_2 \neq 0$ .  $\Delta D = 176.61 > 3.84$  so  $H_0$  rejected. [1 mark] With  $\eta_i = \beta_0 + \beta_1 s_i + \beta_2 t_i + \beta_3 s_i t_i$ : Test 3,  $H_0: \beta_3 = 0$   $H_1: \beta_3 \neq 0$ .  $\Delta D = 0.33 < 3.84$  so  $H_0$  not rejected. [1 mark] So both main effects needed but not the interaction.  $H_1$  must be specified in all tests. [1 mark]
- (2)(iiib)  $\eta_3 = 5.091 1.744 \times 3.4 = -0.839$  [1 mark]  $\hat{\mu_3} = \frac{exp(-0.839)}{1 + exp(-0.839)} = 0.302$  [1 mark] Pearson residual is  $\frac{11/40 0.302}{\sqrt{0.302*(1 0.302)/40}} = -0.372$  [1 mark for 11/40, 1 mark for answer]
- (2)(iiic) odds of Y=1 for thickness 5mm is  $exp(5.091-5\times 1.744)=0.027$  odds of Y=1 for thickness 3.4mm is  $exp(5.091-3.4\times 1.744)=0.432$  [1 mark] for both right So odds ratio is 0.027/0.432=0.0625 [1 mark] log OR is difference in the log odds so equals  $1.6\hat{\beta}_2$  so variance is  $1.6^2\times 0.0446$  [1 mark] 95% CI for logOR is  $log(0.0625)\pm 1.96\times 1.6\times \sqrt{0.047}=(-3.44,-2.11)$ [1 mark]

So 95% CI for OR is (0.03,0.12) and so significant evidence that probability of splitting is not the same for the two tyre thicknesses. [1 mark]

- (2)(iiid) Any suitable that include the same thickness in both linear predictors but different surfaces [2 marks]
  - (3)(i) Controlled variables are considered to have fixed marginal totals, response variables are not. [1 mark]

Minimal model contains helmet\*collision only in linear predictor. [1 mark]

### (3)(ii) Percentages are [1 mark]

| helmet     |         |       | no helmet  |         |       |  |  |
|------------|---------|-------|------------|---------|-------|--|--|
| collision  | serious | minor | collision  | serious | minor |  |  |
| lorry      | 52%     |       | lorry      | 60%     |       |  |  |
| car        | 19%     |       | car        | 46%     |       |  |  |
| pedestrian | 11%     |       | pedestrian | 14%     |       |  |  |

Helmets don't seem to affect the probability of a serious injury in collisions with pedestrians or lorries but they seem to reduce it in collisions with cars. [1 mark]

The probability of a serious injury is lower in collisions with pedestrians compared with cars and in collisions with cars compared to lorries. [1 mark]

# (3)(iii) Terms are

- intercept
- indicator variable for helmet
- 2 indicator variables for collision
- indicator variable for outcome
- 2 interaction indicator variables for collision\*helmet

## [1 mark]

$$n = 12, p = 7 \text{ so } df = 5.$$
 [1 mark]

(3)(iv) 
$$\eta_{ijk} = \beta_0 + \alpha_i + \gamma_j + \tau_{ij} + \delta_k + \theta_{ik}$$
  
with  $\alpha_1 = \gamma_1 = \delta_1 = 0$   
and  $\tau_{1.} = \tau_{.1} = \theta_{1.} = \theta_{.1} = 0$  with the notation correctly defined [2 marks].

| (3)(v) Model           | $ ho({ m v})$ Model                  |            |            | Residual Deviance |         |    |  |  |
|------------------------|--------------------------------------|------------|------------|-------------------|---------|----|--|--|
| A helmet $*$ collision | +outcome                             |            |            |                   | 25.643  | 5  |  |  |
| B helmet*collision     |                                      | 20.751     | 4          |                   |         |    |  |  |
| C helmet*collision     | C helmet*collision+outcome*collision |            |            |                   |         |    |  |  |
| D helmet*collision     | +outcome*                            | helmet+out | come*colli | sion              | 2.318   | 2  |  |  |
|                        |                                      |            |            |                   |         |    |  |  |
| Comparison change      | in df                                | change in  | res dev    | chi-sq            | thresho | ld |  |  |
| A,B                    | 1                                    |            | 4.89       |                   | 3.8     | 34 |  |  |
| A,C                    | 2                                    |            | 17.27      |                   | 5.9     | 99 |  |  |
| C,D                    | 1                                    |            | 6.06       |                   | 3.8     | 34 |  |  |

2

B,D

2 marks for each of the three nested comparisons, all have  $\Delta$  res.dev greater than threshold so final model is chosen.

18.43

5.99

Yes seems consistent. The need for the interaction terms not in the minimal model says that the probability of a serious outcome depends on both helmet and collision type and the particular combination of helmet and collision type. [1 mark]

 $(3)(vi) \ \mu = 54 \times \tfrac{10+17}{(10+17)+(44+20)} \ \textbf{[1 mark for 54, M1 for collapsing across levels of helmet,} \\ \textbf{1 mark for correct values in the fraction]} = 54 \times \tfrac{27}{91} = 16.02198 \ \textbf{[1 mark]}$