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Semester 1 Assessment, 2017

School of Mathematics and Statistics

**MAST30025 Linear Statistical Models**

Writing time: 3 hours

Reading time: 15 minutes

This is NOT an open book exam

This paper consists of 8 pages (including this page)

**Authorised materials:**

- Scientific calculators are permitted, but not graphical calculators.
- Two A4 double-sided handwritten sheets of notes.

**Instructions to Students**

- You must NOT remove this question paper at the conclusion of the examination.
- You should attempt all questions. Marks for individual questions are shown.
- The total number of marks available is 90.

**Instructions to Invigilators**

- Students must NOT remove this question paper at the conclusion of the examination.

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**This paper must not be removed from the examination room**

**Question 1 (9 marks)**

- (a) Show that if  $A$  is a symmetric and idempotent matrix, then  $r(A) = \text{tr}(A)$ .
- (b) Show, without using the above result, that if  $X$  is an  $n \times p$  matrix with  $n > p$ , then  $r(X(X^T X)^c X^T) = r(X)$ .
- (c) Let  $B\mathbf{x} = \mathbf{g}$  be a consistent linear system. Show that  $\mathbf{x} = B^c \mathbf{g}$  is a solution to this system.

**Question 2 (13 marks)** Let

$$\mathbf{y} = \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} \sim MVN \left( \begin{bmatrix} 0 \\ -1 \end{bmatrix}, \begin{bmatrix} 2 & 1 \\ 1 & 3 \end{bmatrix} \right).$$

You are given the following R calculations.

```
> V <- matrix(c(2,1,1,3),2,2)
> P <- eigen(V)$vectors
> P%%diag(sqrt(eigen(V)$values))%%t(P)

      [,1]      [,2]
[1,] 1.3763819 0.3249197
[2,] 0.3249197 1.7013016

> P%%diag(1/sqrt(eigen(V)$values))%%t(P)

      [,1]      [,2]
[1,] 0.7608452 -0.1453085
[2,] -0.1453085 0.6155367
```

- (a) Find two independent standard normal random variables which are linear combinations of  $\mathbf{y}$  elements (and constants).
- (b) Calculate  $E[y_1^2 - 2y_1 y_2]$ .
- (c) Find all quadratic forms in  $\mathbf{y}$  which have a non-central  $\chi^2$  distribution with 2 degrees of freedom.
- (d) Show that  $4y_1^2 - 4y_1 y_2 + y_2^2$  is independent of  $y_1^2 + 6y_1 y_2 + 9y_2^2$ .

**Question 3 (14 marks)** The following data is a sample of 5 random countries. For each country we measure the following variables:

- **logPPgdp**: The logarithm of the 2001 gross domestic product per person in US dollars;
- **logFertility**: The logarithm of the birth rate per 1000 females in the year 2000;
- **Purban**: The percentage of the population which lives in an urban area.

| Name        | logFertility | logPPgdp | Purban |
|-------------|--------------|----------|--------|
| Moldova     | 0.23         | 5.8      | 41     |
| Netherlands | 0.38         | 10.1     | 90     |
| Estonia     | 0.14         | 8.3      | 69     |
| Uganda      | 1.36         | 5.5      | 15     |
| Hungary     | 0.13         | 8.6      | 65     |

We wish to predict **logFertility** using **logPPgdp** and **Purban**, using a linear model  $\mathbf{y} = X\boldsymbol{\beta} + \boldsymbol{\varepsilon}$  with no constant (intercept) term.

You are given the following R calculations.

```
> qt(0.975,1:5)
```

```
[1] 12.706205  4.302653  3.182446  2.776445  2.570582
```

```
> qf(0.95,1,1:5)
```

```
[1] 161.447639 18.512821 10.127964  7.708647  6.607891
```

```
> qf(0.95,2,1:5)
```

```
[1] 199.500000 19.000000  9.552094  6.944272  5.786135
```

- Calculate the least squares estimates of  $\boldsymbol{\beta}$ .
- Calculate and interpret a 95% confidence interval for the parameter associated with **Purban**. You are given the sample variance  $s^2 = 0.0887$ .
- Test for model relevance at a 5% significance level.
- Croatia has a gross domestic product of \$4500 per person, and 58% of its population lives in an urban area. Calculate a 95% prediction interval for its birth rate.

**Question 4 (13 marks)** Consider a full rank linear model  $\mathbf{y} = X\boldsymbol{\beta} + \boldsymbol{\varepsilon}$ . We wish to derive the formula for a prediction interval for the sum of the responses of two independent future observations with predictors  $\mathbf{x}_1$  and  $\mathbf{x}_2$  respectively. This uses the unbiased point estimator  $(\mathbf{x}_1 + \mathbf{x}_2)^T \mathbf{b}$ , where  $\mathbf{b}$  is the least squares estimator of  $\boldsymbol{\beta}$ .

- (a) Calculate the variance of the prediction error of this estimator.
- (b) Show that this estimator is normally distributed.
- (c) Show that this estimator is independent of  $SS_{Res}$ .
- (d) Derive a  $t$ -distributed quantity based on this estimator and state its degrees of freedom.
- (e) Thus write down a formula for a  $100(1 - \alpha)\%$  prediction interval for the sum of two responses.

**Question 5 (14 marks)** Consider the general linear model,  $\mathbf{y} = X\boldsymbol{\beta} + \boldsymbol{\varepsilon}$ . This model may be of full or less than full rank.

- (a) State two methods that can be used to fit this model to data and compare them.
- (b) Explain the difference between a confidence interval and a prediction interval.
- (c) Define Akaike's information criterion and explain why it is useful as a goodness-of-fit measure for model selection.
- (d) Give an advantage and a disadvantage of stepwise selection over forward selection.
- (e) Define interaction between two categorical predictors.
- (f) List two principles of experimental design.
- (g) Explain what a Latin square is and its use in experimental design.

**Question 6 (15 marks)** Data was collected on the world record times for the one-mile run. For males, the records are from the period 1861–2003, and for females, from the period 1967–2003. This data is analysed below.

```
> mile <- read.csv('mile.csv', header=T)
> mile$Gender <- factor(mile$Gender)
> plot(Time ~ Year, data = mile, pch=as.character(Gender))
> imodel <- lm(Time ~ (Year + Gender)^2, data = mile)
> summary(imodel)
```

Call:

```
lm(formula = Time ~ (Year + Gender)^2, data = mile)
```

Residuals:

| Min     | 1Q      | Median  | 3Q     | Max     |
|---------|---------|---------|--------|---------|
| -5.4512 | -1.6160 | -0.1137 | 1.1784 | 13.7265 |

Coefficients:

|                 | Estimate   | Std. Error | t value | Pr(> t )     |
|-----------------|------------|------------|---------|--------------|
| (Intercept)     | 2309.4247  | 202.0583   | 11.429  | < 2e-16 ***  |
| Year            | -1.0337    | 0.1021     | -10.126 | 1.95e-14 *** |
| GenderMale      | -1355.6778 | 203.1441   | -6.673  | 1.03e-08 *** |
| Year:GenderMale | 0.6675     | 0.1027     | 6.502   | 2.00e-08 *** |

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

Residual standard error: 2.989 on 58 degrees of freedom

Multiple R-squared: 0.9663, Adjusted R-squared: 0.9645

F-statistic: 553.8 on 3 and 58 DF, p-value: < 2.2e-16

```
> amodel <- lm(Time ~ Year + Gender, data = mile)
> summary(amodel)
```

Call:

```
lm(formula = Time ~ Year + Gender, data = mile)
```

Residuals:

| Min     | 1Q      | Median  | 3Q     | Max     |
|---------|---------|---------|--------|---------|
| -5.9071 | -2.0988 | -0.1141 | 1.2002 | 13.1863 |

Coefficients:

|             | Estimate   | Std. Error | t value | Pr(> t )   |
|-------------|------------|------------|---------|------------|
| (Intercept) | 1003.00334 | 27.84691   | 36.02   | <2e-16 *** |
| Year        | -0.37364   | 0.01406    | -26.57  | <2e-16 *** |
| GenderMale  | -34.85078  | 1.30099    | -26.79  | <2e-16 *** |

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

Residual standard error: 3.896 on 59 degrees of freedom

Multiple R-squared: 0.9417, Adjusted R-squared: 0.9397  
 F-statistic: 476.3 on 2 and 59 DF, p-value: < 2.2e-16

```
> anova(amodel, imodel)
```

Analysis of Variance Table

Model 1: Time ~ Year + Gender

Model 2: Time ~ (Year + Gender)^2

|   | Res.Df | RSS    | Df | Sum of Sq | F      | Pr(>F)        |
|---|--------|--------|----|-----------|--------|---------------|
| 1 | 59     | 895.62 |    |           |        |               |
| 2 | 58     | 518.03 | 1  | 377.59    | 42.276 | 2.001e-08 *** |

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

```
> linearHypothesis(imodel, c(0,1,0,1), -0.3)
```

Linear hypothesis test

Hypothesis:

Year + Year:GenderMale = - 0.3

Model 1: restricted model

Model 2: Time ~ (Year + Gender)^2

|   | Res.Df | RSS    | Df | Sum of Sq | F      | Pr(>F)        |
|---|--------|--------|----|-----------|--------|---------------|
| 1 | 59     | 850.63 |    |           |        |               |
| 2 | 58     | 518.03 | 1  | 332.6     | 37.238 | 9.236e-08 *** |

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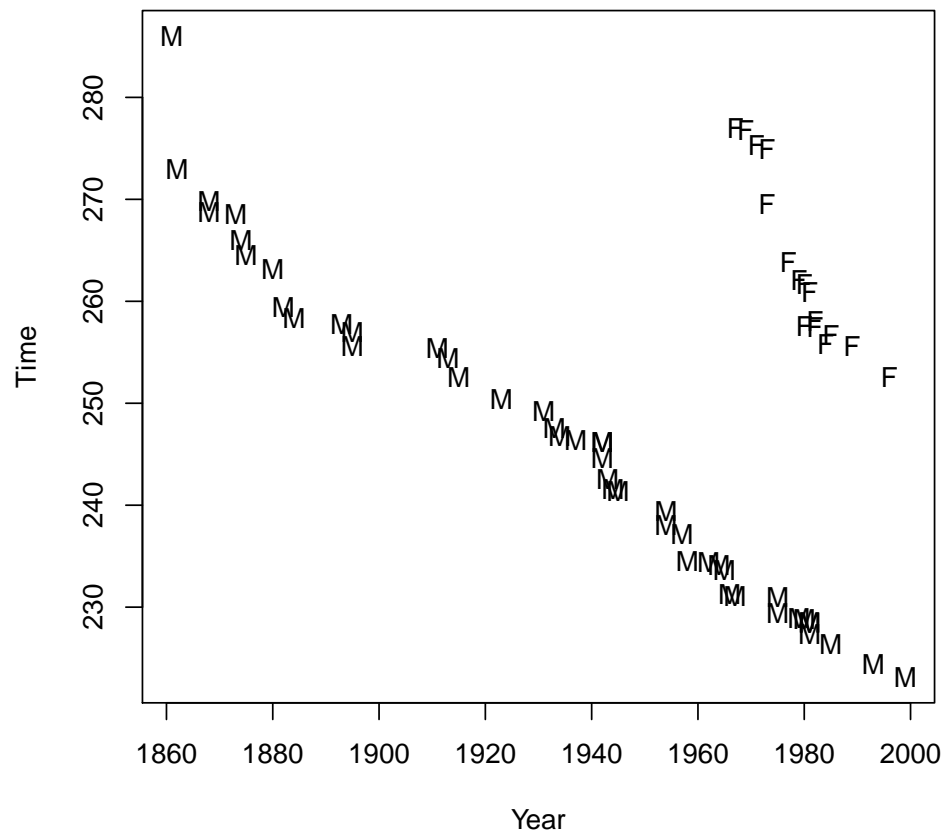
Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

```
> qt(c(0.95,0.975,0.99,0.995), df=58)
```

```
[1] 1.671553 2.001717 2.392377 2.663287
```

```
> qt(c(0.95,0.975,0.99,0.995), df=59)
```

```
[1] 1.671093 2.000995 2.391229 2.661759
```



- Do you think that this data satisfies the linear model assumptions? Explain.
- What are the covariates used in the `imodel` object (in the context of the study)?
- What is being tested in the `anova` function call? What do you conclude from the results?
- Write down the final fitted models for the male and female records.
- Calculate a point estimate for the year when the female world record will equal the male world record. Do you expect this estimate to be accurate? Why or why not?
- Calculate a 95% confidence interval for the amount by which the gap between the male and female world records narrow every year.
- What is the hypothesis being tested in the `linearHypothesis` function call? What do you conclude from the output?

**Question 7 (12 marks)**

- (a) Randomisation eliminates confounding from both known and unknown factors. Explain why it is additionally advantageous to use blocking for some confounding factors.
- (b) A study is to be conducted to evaluate the effect of a drug on brain function. The evaluation consists of measuring the response of a particular part of the brain using an MRI scan. The drug is prescribed in doses of 1, 2 and 5 milligrams. Funding allows only 24 observations to be taken in the current study.

Explain how control might be used in a design of this experiment.

- (c) For the scenario above, explain how replication might be used in a design of this experiment.
- (d) For a complete block design with  $b$  blocks and  $k$  treatments, the reduced design matrix  $X_{2|1}$  satisfies

$$X_{2|1}^T X_{2|1} = b \left[ I_k - \frac{1}{k} J_k \right],$$

where  $I_k$  is the  $k \times k$  identity matrix and  $J_k$  is the  $k \times k$  matrix with all elements equal to 1. Show that

$$(X_{2|1}^T X_{2|1})^c = \frac{1}{b} I_k.$$

- (e) For the above complete block design, show that if a quantity  $\mathbf{t}^T \boldsymbol{\tau}$  involving only the treatment parameters is estimable, then it must be a treatment contrast. (That is,  $\mathbf{t}^T \mathbf{1} = 0$ .)

**End of Exam—Total Available Marks = 90.**





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