

**UNIVERSITY OF NEW SOUTH WALES  
SCHOOL OF MATHEMATICS AND STATISTICS**

**MATH2831/2931 Linear Models  
Assignment Two**

**Note:** This is a group assignment and is due on Friday's lecture in Week 8.

Names (Print): \_\_\_\_\_

I (We) declare that this assessment item is my (our) own work, except where acknowledged, and has not been submitted for academic credit elsewhere, and acknowledge that the assessor of this item may, for the purpose of assessing this item:

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I (We) certify that I (We) have read and understood the University Rules in respect of Student Academic Misconduct.

Signed: \_\_\_\_\_

Date: \_\_\_\_\_

Please follow the instructions below for completing the assignment (worth 10%).

1. You can work in groups of up to 3 people and submit a single (typeset) assignment (with the names of these 3 people). All members of the group will get the same marks.
2. You can split the work amongst the members of the group in any way you like, but free-riders should note that all of the assignment questions are examinable and potentially on the final exam. Thus, it is in everybody's interest to understand the assignment questions and be able to solve them.
3. Individuals who do not like working in groups are still free to submit an individual assignment.
4. To properly typeset the assignment, work with the package <http://www.latex-project.org/>. There is a large amount of information available on Latex at this URL including the installation guides for the compiler (miktex) and the editors (free of charge). You can use either Lyx <http://www.lyx.org/> or Texnic <http://www.texniccenter.org/>. Both are free latex editors that you can easily install and get started with. Once you have completed installation, you may use the latex template provided in the Assignment folder on Moodle.

Use the R-output given to answer questions 1,2,3.

1. The data is partly taken from the *Energy efficiency Data Set* on the *UCI Machine Learning Repository*<sup>1</sup>. The study performed energy analysis using 12 different building shapes. The buildings differ with respect to the glazing area, the glazing area distribution, and the orientation, amongst other parameters. The response in this data set is the `Cooling_Load` and `Heating_Load`.

To study the energy required in cooling the building, we fit the following multiple linear regression model using the predictors `Relative_Compactness`, `Surface_Area` and `Wall_Area`.

```
> summary.lm(cooling)
```

Call:

```
lm(formula = Cooling_Load ~ Relative_Compactness + Surface_Area +  
    Wall_Area, data = energyMod)
```

Residuals:

	Min	1Q	Median	3Q	Max
	-10.6007	-3.1655	-0.7148	2.7230	11.4541

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	-1.440e+03	1.835e+02	-7.847	4.37e-14 ***
Relative_Compactness	8.184e+02	1.039e+02	7.873	3.65e-14 ***
Surface_Area	1.358e+00	1.692e-01	8.028	1.25e-14 ***
Wall_Area	-1.017e-01	1.965e-02	-5.179	3.63e-07 ***

---

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

Residual standard error: 4.427 on 380 degrees of freedom

Multiple R-squared: 0.3508, Adjusted R-squared: 0.3457

F-statistic: 68.44 on 3 and 380 DF, p-value: < 2.2e-16

- (a) State the value of the  $F$  statistic used to test the hypothesis that  $\beta_1 = \beta_2 = \beta_3 = 0$  versus  $\beta_1 \neq 0$  or  $\beta_2 \neq 0$  or  $\beta_3 \neq 0$ . State the conclusion of the test under a 1% level.
- (b) Construct an 95% confidence interval for the parameter  $\beta_3$  associated with `Wall_Area`.
- (c) How many observations are there in this data set?

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<sup>1</sup>URL: <http://archive.ics.uci.edu/ml/datasets/Energy+efficiency>

2. The ANOVA table of the fitted model is given below

```
> anova(cooling)
Analysis of Variance Table

Response: Cooling_Load

          Df Sum Sq Mean Sq F value    Pr(>F)
Relative_Compactness  1 2260.5  2260.52 115.358 < 2.2e-16 ***
Surface_Area         1 1237.5  1237.47  63.150  2.20e-14 ***
Wall_Area            1  525.5   525.55  26.819  3.63e-07 ***
Residuals           380 7446.4    19.60
---
Signif. codes:  0 *** 0.001 ** 0.01 * 0.05 . 0.1 1
>
```

- (a) Compute from the above table, the value of the  $F$  statistic used question 1 (a).
- (b) Conduct the appropriate sequential  $F$  test under a 5% level of significance to test whether a model containing all the predictors is preferred over a model with **Relative\_Compactness** as the predictors.
3. Forward selection method was applied on to the dataset (In this case, we use also **Heating\_Load** as an predictor).

```
null<-lm(Cooling_Load~1,data = energyMod)
> full<-lm(Cooling_Load~.,data = energyMod)
> step(null,scope= list(lower = null, upper = full),
direction = 'forward',test = 'F', k = 0.001)
```

```
Start:  AIC=1304.39
Cooling_Load ~ 1
```

	Df	Sum of Sq	RSS	AIC	F value	Pr(>F)
+ Heating_Load	1	8828.8	2641.1	740.48	1276.9559	< 2e-16 ***
+ Wall_Area	1	2554.1	8915.9	1207.66	109.4286	< 2e-16 ***
+ Surface_Area	1	2476.5	8993.4	1210.99	105.1927	< 2e-16 ***
+ Relative_Compactness	1	2260.5	9209.4	1220.10	93.7649	< 2e-16 ***
+ Roof_Area	1	95.0	11374.9	1301.19	3.1913	0.07482 .
<none>			11469.9	1304.39		

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

```
Step:  AIC=740.48
```

Cooling\_Load ~ Heating\_Load

	Df	Sum of Sq	RSS	AIC	F value	Pr(>F)	
+ Relative_Compactness	1	49.495	2591.6	733.21	7.2763	0.007297	**
+ Surface_Area	1	48.548	2592.6	733.35	7.1346	0.007885	**
+ Roof_Area	1	44.888	2596.2	733.89	6.5874	0.010652	*
+ Wall_Area	1	7.618	2633.5	739.37	1.1021	0.294469	
<none>			2641.1	740.48			

---

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

Step: AIC=733.21

Cooling\_Load ~ Heating\_Load + Relative\_Compactness

	Df	Sum of Sq	RSS	AIC	F value	Pr(>F)
+ Roof_Area	1	9.3064	2582.3	731.83	1.3695	0.2426
+ Wall_Area	1	8.0475	2583.6	732.02	1.1836	0.2773
+ Surface_Area	1	0.5994	2591.0	733.12	0.0879	0.7670
<none>			2591.6	733.21		

Step: AIC=731.83

Cooling\_Load ~ Heating\_Load + Relative\_Compactness + Roof\_Area

	Df	Sum of Sq	RSS	AIC	F value	Pr(>F)
+ Surface_Area	1	16.422	2565.9	729.38	2.4256	0.1202
+ Wall_Area	1	16.422	2565.9	729.38	2.4256	0.1202
<none>			2582.3	731.83		

Step: AIC=729.38

Cooling\_Load ~ Heating\_Load + Relative\_Compactness + Roof\_Area +  
Surface\_Area

	Df	Sum of Sq	RSS	AIC	F value	Pr(>F)
<none>			2565.9	729.38		

Call:

```
lm(formula = Cooling_Load ~ Heating_Load + Relative_Compactness +  
Roof_Area + Surface_Area, data = energyMod)
```

Coefficients:

(Intercept)	Heating_Load	Relative_Compactness
-173.36976	0.76691	100.15325
Roof_Area	Surface_Area	
0.04591	0.15281	

- (a) Perform the forward selection procedure using the sequential  $F$ -test as the selection rule. (i) What is the final model under a 5% level? (ii) What about under a 10% level?
  - (b) What is the sum of square regression for the model with all the predictors except **Wall\_Area**.
  - (c) Conduct an appropriate hypothesis test to test under a 5% level, whether the model including only **Heating\_Load** and **Relative\_Compactness** is preferred over the model including **Heating\_Load**, **Relative\_Compactness**, **Roof\_Area** and **Surface\_Area**.
4. Observations  $(x_i, y_i)$  for  $i = 1, \dots, n$  are made according to the model

$$y_i = \alpha + \beta x_i + \epsilon_i$$

where  $x_i, \dots, x_n$  are fixed constants and  $\epsilon_i$  are i.i.d  $\mathcal{N}(0, \sigma^2)$ . Suppose the model is then re-parametrized as

$$y_i = \alpha' + \beta'(x_i - \bar{x}) + \epsilon_i$$

let  $\hat{\alpha}$  and  $\hat{\beta}$  denotes the MLEs of  $\alpha$  and  $\beta$ , respectively, and  $\hat{\alpha}'$  and  $\hat{\beta}'$  denote the MLEs of  $\alpha'$  and  $\beta'$ , respectively

- (a) Show that  $\hat{\beta}' = \hat{\beta}$
  - (b) Show that  $\hat{\alpha}' \neq \hat{\alpha}$
  - (c) Show that  $\hat{\alpha}'$  and  $\hat{\beta}'$  are uncorrelated.
5. If  $X$  is the design matrix, show that

$$SS_{reg} = y^T(X(X^T X)^{-1}X^T - X_2(X_2^T X_2)^{-1}X_2^T)y.$$

where  $X_2$  is a  $n \times 1$ -matrix with entries equal to one.

(Hint: if  $\hat{y}$  denotes the  $n \times 1$  vector of fitted values, use the fact that

$$\sum_{i=1}^n (\hat{y}_i - \bar{y})^2 = \sum_{i=1}^n \hat{y}_i^2 - \frac{(\sum_{i=1}^n y_i)^2}{n}$$

and show that

$$\sum_{i=1}^n \hat{y}_i^2 = \hat{y}^T \hat{y} = y^T X(X^T X)^{-1}X^T y.$$

6. **(MATH2931 only)** The **Sherman-Morrison** formula, a special case of *matrix blockwise inversion*, states that for a matrix  $A$  and vectors  $u$  and  $v$  of appropriate size

$$(A + uv^T)^{-1} = A^{-1} - \frac{A^{-1}uv^T A^{-1}}{1 + v^T A^{-1}u}, \quad v^T A^{-1}u \neq -1.$$

Prove the Sherman-Morrison formula and verify the formula with a numerical example in which  $A$  is a  $2 \times 2$  matrix.

7. **(MATH2931 only)** Let  $H = X(X^T X)^{-1} X^T$  be the hat matrix corresponding to the  $n \times p$  design matrix with predictors

$$X = \begin{bmatrix} 1 & x_{1,1} & x_{1,2} & \cdots & x_{1,p-1} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ 1 & x_{n-1,1} & x_{n-1,2} & \cdots & x_{n-1,p-1} \\ 1 & x_{n,1} & x_{n,2} & \cdots & x_{n,p-1} \end{bmatrix} = \begin{bmatrix} x_1^T \\ x_2^T \\ \vdots \\ x_n^T \end{bmatrix}$$

where  $x_n = (1, x_{n,1}, \dots, x_{n,p-1})^T$ . Assuming a full rank linear model, we can write the fitted values as  $\hat{y} = Hy$ . Let  $\hat{y}_{-i}$  be the fitted values for a model fitted by omitting the  $i$ -th observation  $y_i$

- (a) Prove that for the  $i$ -th PRESS residual  $y_i - \hat{y}_{i,-i}$ , we have

$$y_i - \hat{y}_{i,-i} = \frac{y_i - \hat{y}_i}{1 - H_{ii}}.$$

- (b) In no more than three sentences, explain the significance of the PRESS statistic  $\sum_{i=1}^n (y_i - \hat{y}_{i,-i})^2$  and the identity above.

8. **(MATH2931 only)** Assume that  $\beta_0, \dots, \beta_k = 0$ , derive the distribution of  $\frac{SS_{reg}}{\sigma^2}$ .