

NAME (PRINT): \_\_\_\_\_  
Last/Surname First /Given Name

STUDENT #: \_\_\_\_\_ SIGNATURE: \_\_\_\_\_

**UNIVERSITY OF TORONTO MISSISSAUGA  
DECEMBER 2014 FINAL EXAMINATION  
STA302H5F  
Regression Analysis  
Jerry Brunner  
Duration - 3 hours**

**Aids: Calculator Model(s): Any calculator is okay. Formula sheet will be supplied**

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Qn. #	Value	Score
1	5	
2	5	
3	6	
4	3	
5	8	
6	15	
7	3	
8	10	
9	17	
10	13	
11	15	
Total = 100 Points		

Unless otherwise indicated, the questions in this exam refer to the general linear model on the formula sheet, with  $\mathbf{X}$  an  $n \times (k + 1)$  matrix of fixed, observable constants. The columns of the  $\mathbf{X}$  matrix are linearly independent.

*5 points*

1. Show that if the columns of  $\mathbf{X}$  are linearly independent, the matrix  $\mathbf{X}'\mathbf{X}$  is positive definite. It may help to notice that in the definitions on the formula sheet,  $\mathbf{v}$  is  $(k + 1) \times 1$ , so that  $\mathbf{Z} = \mathbf{X}\mathbf{v}$  is  $n \times 1$ .

5 points

2. Let the random vector  $\mathbf{Y}$  have expected value  $\boldsymbol{\mu}$  and variance-covariance matrix  $\boldsymbol{\Sigma}$ , and let  $\mathbf{A}$  and  $\mathbf{B}$  be matrices of constants. Prove that  $C(\mathbf{A}\mathbf{Y}, \mathbf{B}\mathbf{Y}) = \mathbf{A}\boldsymbol{\Sigma}\mathbf{B}'$ .

6 points

3. Now for linear regression, locate the matrices  $\mathbf{A}$  and  $\mathbf{B}$  on the formula sheet and use the last result to show  $C(\hat{\boldsymbol{\epsilon}}, \hat{\boldsymbol{\beta}}) = \mathbf{0}$ .

3 points

4. Why does  $C(\hat{\boldsymbol{\epsilon}}, \hat{\boldsymbol{\beta}}) = \mathbf{0}$  establish that  $\hat{\boldsymbol{\beta}}$  and  $\hat{\boldsymbol{\epsilon}}$  are independent?

8 points

5. For the usual linear regression model,

(a) Calculate  $\mathbf{X}'\hat{\boldsymbol{\epsilon}}$  and simplify.

(b) Without going all the way back to the normal equations, why does the last result show that if a regression model has an intercept, then the residuals must add up to zero?

(c) Your answer to Question 5a also shows that geometrically, the vector of residuals is at right angles to every column vector of the  $\mathbf{X}$  matrix. So, it should be at right angles to any linear combination of those column vectors. A general way to represent such a linear combination is  $\mathbf{X}\mathbf{b}$ , where  $\mathbf{b}$  is a  $(k + 1) \times 1$  vector that could be random or constant. Show  $(\mathbf{X}\mathbf{b})'\hat{\boldsymbol{\epsilon}} = 0$ .

Don't forget this result. It will make your job easier on the next question.

*continued on page 5*

15 points

6. Show that  $(\mathbf{Y} - \mathbf{X}\beta)'(\mathbf{Y} - \mathbf{X}\beta) = \hat{\boldsymbol{\epsilon}}'\hat{\boldsymbol{\epsilon}} + (\hat{\boldsymbol{\beta}} - \boldsymbol{\beta})'(\mathbf{X}'\mathbf{X})(\hat{\boldsymbol{\beta}} - \boldsymbol{\beta})$ .

3 points

7. Why does the result above tell you that  $(\mathbf{Y} - \mathbf{X}\beta)'(\mathbf{Y} - \mathbf{X}\beta)$  is minimal when  $\boldsymbol{\beta} = \hat{\boldsymbol{\beta}}$ ?

10 points

8. In the general linear regression model, let  $\text{cov}(\boldsymbol{\epsilon}) = \sigma^2 \mathbf{V}$ , where  $\mathbf{V}$  is a *known* symmetric and positive definite matrix. As usual,  $\sigma^2$  is an unknown constant. Suppose we try to estimate  $\boldsymbol{\beta}$  with  $\hat{\boldsymbol{\beta}}_w = (\mathbf{X}'\mathbf{V}^{-1}\mathbf{X})^{-1}\mathbf{X}'\mathbf{V}^{-1}\mathbf{Y}$ .

(a) Is  $\hat{\boldsymbol{\beta}}_w$  an unbiased estimator of  $\boldsymbol{\beta}$ ? Answer Yes or No and prove your answer.

(b) If  $\boldsymbol{\epsilon} \sim N_n(\mathbf{0}, \sigma^2 \mathbf{V})$ , what is the distribution of  $\hat{\boldsymbol{\beta}}_w$ ? Use the formula sheet, show your work and simplify.

17 points

9. Question 6 established that  $(\mathbf{Y} - \mathbf{X}\boldsymbol{\beta})'(\mathbf{Y} - \mathbf{X}\boldsymbol{\beta}) = \hat{\boldsymbol{\epsilon}}'\hat{\boldsymbol{\epsilon}} + (\hat{\boldsymbol{\beta}} - \boldsymbol{\beta})'(\mathbf{X}'\mathbf{X})(\hat{\boldsymbol{\beta}} - \boldsymbol{\beta})$ .

(a) What is the distribution of  $\frac{1}{\sigma^2}(\mathbf{Y} - \mathbf{X}\boldsymbol{\beta})'(\mathbf{Y} - \mathbf{X}\boldsymbol{\beta})$ ? Prove it, citing any facts from the formula sheet as you use them. Include the degrees of freedom in your answer.

(b) What is the distribution of  $\frac{1}{\sigma^2}(\hat{\boldsymbol{\beta}} - \boldsymbol{\beta})'(\mathbf{X}'\mathbf{X})(\hat{\boldsymbol{\beta}} - \boldsymbol{\beta})$ ? Prove it, citing any facts from the formula sheet as you use them. Include the degrees of freedom in your answer.

(c) How do you know  $\frac{1}{\sigma^2}\widehat{\boldsymbol{\epsilon}}'\widehat{\boldsymbol{\epsilon}}$  and  $\frac{1}{\sigma^2}(\widehat{\boldsymbol{\beta}} - \boldsymbol{\beta})'(\mathbf{X}'\mathbf{X})(\widehat{\boldsymbol{\beta}} - \boldsymbol{\beta})$  are independent?

(d) Finally (this was the goal), show  $\frac{1}{\sigma^2}\widehat{\boldsymbol{\epsilon}}'\widehat{\boldsymbol{\epsilon}} \sim \chi^2(n - k - 1)$ .



13 points

10. In a study comparing the effectiveness of different weight loss diets, volunteers were randomly assigned to one of two diets ( $A$  or  $B$ ) or put on a waiting list and advised to lose weight on their own. Participants were weighed before and after 6 months of participation in the program (or 6 months of being on the waiting list). The response variable is weight *loss*. The explanatory variables are age and treatment group.
- (a) Write the regression equation. Your model should have *parallel regression lines* for the different treatment groups. Please use  $x$  for age. You don't have to say how your dummy variables are defined. You'll do that in the next part.
- (b) Make a table with three rows, showing how you would set up indicator dummy variables for treatment group. Give  $E(Y|x)$  in the last column.
- (c) In terms of  $\beta$  values, what null hypothesis would you test to find out whether, allowing for age, the three diets (including Wait List) differ in their effectiveness?
- (d) In terms of  $\beta$  values, what null hypothesis would you test to find out whether, allowing for age, diets  $A$  and  $B$  differ in their effectiveness?
- (e) In terms of  $\beta$  values, what null hypothesis would you test to find out whether 25 year old participants who spend six months on the Wait list "diet" have any average tendency to gain or lose weight? Remember,  $Y$  is weight loss.

11. (15 points) This question is based on a study of low birth weight in babies. The dependent variable is `low`, which equals 1 if the baby is less than 2.5 kg (low birth weight), and zero otherwise. The independent variables are mother's weight (`lwt`), an indicator variable for smoking (`smoke`), and mother's race.

```
> # Birth weight: MASS package must be loaded
> attach(birthwt)
> # low is low birth weight, lwt is mother's weight, smoke is indicator for smoking
> race = factor(race, labels = c("White", "Black", "Other"))
> contrasts(race)
      Black Other
White    0     0
Black    1     0
Other    0     1
> baby = glm(low ~ lwt+smoke+race, family=binomial); summary(baby)
```

```
Call:
glm(formula = low ~ lwt + smoke + race, family = binomial)
```

```
Deviance Residuals:
    Min       1Q   Median       3Q      Max
-1.5278 -0.9053 -0.5863  1.2878  2.0364
```

```
Coefficients:
            Estimate Std. Error z value Pr(>|z|)
(Intercept) -0.10922   0.88211  -0.124  0.90146
lwt          -0.01326   0.00631  -2.101  0.03562 *
smoke         1.06001   0.37832   2.802  0.00508 **
raceBlack    1.29009   0.51087   2.525  0.01156 *
raceOther    0.97052   0.41224   2.354  0.01856 *
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
Null deviance: 234.67 on 188 degrees of freedom
Residual deviance: 215.01 on 184 degrees of freedom
AIC: 225.01
```

```
Number of Fisher Scoring iterations: 4
```

```
> mod1 = glm(low ~ lwt+smoke, family=binomial)
> anova(mod1,baby,test='Chisq')
```

```
Analysis of Deviance Table

Model 1: low ~ lwt + smoke
Model 2: low ~ lwt + smoke + race
  Resid. Df Resid. Dev Df Deviance Pr(>Chi)
1         186      224.34
2         184      215.01  2    9.326 0.009438 **
---
```

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

```
> mod2 = glm(low ~ 1, family=binomial)
> anova(mod2,baby,test='Chisq')
```

```
Analysis of Deviance Table

Model 1: low ~ 1
Model 2: low ~ lwt + smoke + race
  Resid. Df Resid. Dev Df Deviance Pr(>Chi)
1         188      234.67
2         184      215.01  4   19.657 0.0005835 ***
---
```

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

- (a) Controlling for race and weight, the estimated odds of a low birth weight baby are \_\_\_\_\_ times as great for a mother who smokes. The answer is a number. Write your answer in the space below. **Circle your answer.**
- (b) Controlling for smoking and weight, the estimated odds of a low birth weight baby are \_\_\_\_\_ times as great for a Black mother as for a White mother. The answer is a number. Write your answer in the space below. **Circle your answer.**
- (c) We want to know whether, controlling for smoking and weight, race is related to the chances of having a low birth weight baby.
- Give the value of test statistic (not the  $p$ -value). The answer is a number.
  - Do you reject the null hypothesis at  $\alpha = 0.05$ ? Answer Yes or No.
  - Controlling for smoking and weight, is there evidence that race is related to the chances of having a low birth weight baby? Answer Yes or No.
- (d) We want to know whether, controlling for smoking and weight, White and Other mothers differ in their chances of having a low birth weight baby.
- Give the value of test statistic (not the  $p$ -value). The answer is a number.
  - Do you reject the null hypothesis at  $\alpha = 0.05$ ? Answer Yes or No.
  - In plain, non-statistical language, what do you conclude?
- (e) Estimate the probability of a low birth weight baby for a 130-pound, non-smoking White mother. The answer is a number. **Circle your answer.**