

**Bios 312:
Modern Regression Analysis**

**Final Examination
April 28, 2011**

Name: _____

Instructions: Please provide concise answers to all questions. Questions are of varying levels of difficulty, so you may find it advantageous to skip questions you find especially difficult, and return to these questions at the end of the exam.

You are allowed up to six (6) pages of your own notes to assist you when taking the exam.

You may use a calculator to assist with arithmetic. When making intermediate calculations, always use at least four significant digits; report at least three significant digits.

If you come to a problem that you believe cannot be answered without making additional assumptions, clearly state the reasonable assumption that you make, and proceed.

Please adhere to the following pledge. If you are unable to truthfully sign the pledge for any reason, turn in your paper unsigned and discuss the circumstances with the instructor.

PLEDGE: On my honor, I have neither given nor received unauthorized aid on this examination

This exam consists of

- 11 total pages including the Appendix of Results
- There are 135 total points
 - Question 1: parts (a) – (e), 25 points
 - Question 2: parts (a) – (d), 20 points
 - Question 3: parts (a) – (m), 65 points
 - Question 4: parts (a) – (b), 10 points
 - Question 5: parts (a) – (e), 15 points

Question 1 (25 points total, 5 points each). Consider the following univariate regression models exploring the association between survival and body mass index (BMI). BMI is a continuous measure of weight divide by height-squared ($\text{BMI} = \text{kg} / \text{m}^2$).

For each model given below, provide an interpretation of the slope parameter in terms of which summary measure is compared across groups. Indicate whether such an analysis would be appropriate in this dataset. Note that while some subjects are censored in this dataset, no subject is censored before 4 years time. (You do not have any results for the following analyses. Just indicate what you would be examining.)

1.a) A linear regression of BMI (response) on a variable indicating that death occurred within 4 years (predictor).

1.b) A linear regression of a variable indicating that death occurred within 4 years (response) on BMI (predictor).

1.c) A logistic regression of a variable indicating that death occurred within 4 years (response) on BMI (predictor).

1.d) A proportional hazards regression of the observation time and a variable indicating that death occurred within 4 years (response) on BMI (predictor).

1.e) Very briefly indicate the relative advantages and disadvantages of these four approaches.

Question 2 (20 points total, 5 points each). Each of following logistic regression models examine the association between some function of BMI and death occurring within 4 years. In Model 2.1, BMI is included as a linear term. In 2.2, BMI is modeled using a linear and a quadratic term. Model 2.3 uses dummy variables to indicate medium and high BMI using cut-points of 18 and 30. The reference group for model 2.3 is low BMI (defined to be BMI of 18 or lower).

More specifically, the models are given as:

$$\text{Model 2.1: } \log(\text{odds of death}) = \alpha_0 + \alpha_1 * \text{bmi}$$

$$\text{Model 2.2: } \log(\text{odds of death}) = \beta_0 + \beta_1 * \text{bmi} + \beta_2 * \text{bmi}^2$$

$$\text{Model 2.3: } \log(\text{odds of death}) = \gamma_0 + \gamma_1 * \text{bmi_middle} + \gamma_2 * \text{bmi_high}$$

$\text{bmi_middle} = 1$ if BMI is within the range (18,30], and 0 otherwise

$\text{bmi_high} = 1$ if BMI is above 30, and 0 otherwise

Each of these models are fit in Appendix A. For each model, output is provided for the log-odds of death (obtained by the logit command) and the odds of death (obtained by the logistic command). Use this output to answer the following questions.

2.1) Using the results from Model 2.1, is there an association between the odds of death and BMI? Justify your answer.

2.2) Using the results from Model 2.2, is there an association between the odds of death and BMI? Justify your answer.

2.3) Using the results from Model 2.3, is there an association between the odds of death and BMI? Justify your answer.

2.4) Compared the results from each of the models. Based on the given output, is there any evidence to suggest that there is a non-linear association between the odds of death and BMI? Why or why not?

Question 3(65 points total, 5 points each). Appendix B contains the results of a logistic regression performed on a variable indicating death observed with 4 years on BMI, age, gender, the BMI-gender interaction, and the BMI-age interaction. Use the output in Appendix B to answer the following questions.

3.a) What is the estimated odds of death within 4 years for a 60 year old female with a BMI of 30 kg/m²?

3.b) What is the estimated probability of death within 4 years for a 60 year old female with a BMI of 30 kg/m²?

3.c) What is the estimated odds of death within 4 years for a 61 year old female with a BMI of 30 kg/m²?

3.d) What is the estimated odds of death within 4 years for a 60 year old male with a BMI of 30 kg/m²?

3.e) What is the interpretation of the intercept in the regression model obtained using the logit command?

3.f) What is the interpretation of the slope parameter for BMI obtained using the logit command?

3.g) What is the interpretation of the slope parameter for age obtained using the logit command?

3.h) What is the interpretation of the slope parameter for gender obtained using the logit command?

3.i) What is the interpretation of the slope parameter for the gender-BMI interaction obtained using the logit command?

3.j) Using this model, explain how you would test if BMI is significantly associated with death within 4 years. Give the null and alternative hypothesis for this test.

3.k) Using this model, explain how you would test if gender is significantly associated with death within 4 years. Give the null and alternative hypothesis for this test.

3.l) Suppose we fit a logistic regression model of death with 4 years on BMI and age in just females. What would the estimate of the intercept, BMI slope, and age slope in such a model?

3.m) Suppose we fit a logistic regression model of death with 4 years on BMI and age in just males. What would the estimate of the intercept, BMI slope, and age slope in such a model?

Question 4 (10 points total, 5 points each). Suppose we are interested in exploring the association between systolic blood pressure and age and gender. Consider two possible study designs:

- Study A: Gather a single blood pressure measurement on 5,000 independent subjects of both sexes between the ages of 60 and 80
- Study B: Gather five measurements made one year apart on each of 1,000 independent subjects of both sexes between the age of 60 and 80

4.a) Is Study A or Study B more likely to provide more statistical precision to assess an association between blood pressure and age? Explain why.

4.b) Is Study A or Study B more likely to provide more statistical precision to assess an association between blood pressure and gender? Explain why.

Question 5 (15 points total, 3 points each). The Scholastic Aptitude Test (SAT) is a standardized test that many colleges and universities use in evaluating undergraduate students for admission. Assume that the SAT is rigorously designed and evaluated so that, each year, scores follow a Normal distribution with a mean of 500 points and standard deviation of 100 points. An investigator has access to a random sample of SAT scores from North Carolina and Tennessee. For each score, the investigator also has data on whether the test was taken in the junior or senior year of high school. They want to evaluate the impact of state and year on SAT scores using the following linear regression model:

$$E[\text{SAT} \mid \text{state}, \text{year}] = \beta_0 + \beta_1 * \text{state} + \beta_2 * \text{year} + \beta_3 * \text{state} * \text{year}$$

where state=1 if Tennessee and state=0 if North Carolina,
year=1 if taken in senior year or year=0 if taken in junior year

For each of the following assumptions of classical linear regression, indicate whether or not you expect the assumption to hold for this analysis, and if this assumption is necessary to make statistical inference about association, means in groups, and/or prediction (forecasting) of new individual observations. If you believe an assumption might not hold, briefly explain how you would evaluate the assumption.

5.1 All of the observations are independent

5.2 The parameter estimates (the β s) follow a Normal distribution

5.3 Constant variance across groups (homoskedasticity)

5.4 The mean model has been appropriately specified

5.5 The residuals follow a Normal distribution

Appendix for Question 2: Different dose-response models for BMI (Appendix A)

```
. gen death4 = 0
. replace death4 = 1 if obstime <=4 & death==1
(238 real changes made)
```

```
. gen bmi_middle = 0
. replace bmi_middle = 1 if bmi <=30 & bmi>18
(1983 real changes made)
```

```
. gen bmi_high = 0
. replace bmi_high = 1 if bmi >30
(495 real changes made)
```

```
. gen bmi_sqr = bmi*bmi
```

Model 2.1: BMI modeled as a linear function

```
. logit death4 bmi
```

```
Logistic regression                Number of obs   =      2500
                                   LR chi2(1)       =         1.26
                                   Prob > chi2        =      0.2622
Log likelihood = -785.38747        Pseudo R2      =      0.0008
```

death4	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
bmi	-.0162804	.0146713	-1.11	0.267	-.0450356 .0124747
_cons	-1.81808	.394575	-4.61	0.000	-2.591432 -1.044727

```
. logistic death4 bmi
```

```
Logistic regression                Number of obs   =      2500
                                   LR chi2(1)       =         1.26
                                   Prob > chi2        =      0.2622
Log likelihood = -785.38747        Pseudo R2      =      0.0008
```

death4	Odds Ratio	Std. Err.	z	P> z	[95% Conf. Interval]
bmi	.9838514	.0144344	-1.11	0.267	.9559634 1.012553

Model 2.2: BMI modeled as a quadratic function

```
. logit death4 bmi bmi_sqr
```

```
Logistic regression                Number of obs   =      2500
                                   LR chi2(2)        =         5.24
                                   Prob > chi2         =      0.0727
Log likelihood = -783.39493        Pseudo R2      =      0.0033
```

death4	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
bmi	-.2007815	.0889987	-2.26	0.024	-.3752158 - .0263472
bmi_sqr	.0031772	.0014965	2.12	0.034	.0002441 .0061103
_cons	.7648577	1.293666	0.59	0.554	-1.770681 3.300397

```
. logistic death4 bmi bmi_sqr
```

```
Logistic regression                Number of obs   =      2500
                                   LR chi2(2)        =         5.24
                                   Prob > chi2         =      0.0727
Log likelihood = -783.39493        Pseudo R2      =      0.0033
```

death4	Odds Ratio	Std. Err.	z	P> z	[95% Conf. Interval]
bmi	.8180912	.0728091	-2.26	0.024	.687141 .9739969
bmi_sqr	1.003182	.0015013	2.12	0.034	1.000244 1.006129

```
. test bmi+bmi_sqr=0
```

```
( 1) [death4]bmi + [death4]bmi_sqr = 0
      chi2( 1) =      5.10
      Prob > chi2 =    0.0240
```

```
. test (bmi=0) (bmi_sqr=0)
```

```
( 1) [death4]bmi = 0
( 2) [death4]bmi_sqr = 0
      chi2( 2) =      5.59
      Prob > chi2 =    0.0611
```

Model 2.3: BMI modeled using indicator variables

```
. logit death4 bmi_middle bmi_high

Logistic regression                Number of obs   =      2500
                                   LR chi2(2)        =      8.44
                                   Prob > chi2         =      0.0147
Log likelihood = -781.79504        Pseudo R2       =      0.0054
```

death4	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
bmi_middle	-1.235024	.4846175	-2.55	0.011	-2.184857 - .2851912
bmi_high	-1.535122	.5083081	-3.02	0.003	-2.531388 - .5388565
_cons	-.9808292	.4787136	-2.05	0.040	-1.919091 - .0425679

```
. logistic death4 bmi_middle bmi_high

Logistic regression                Number of obs   =      2500
                                   LR chi2(2)        =      8.44
                                   Prob > chi2         =      0.0147
Log likelihood = -781.79504        Pseudo R2       =      0.0054
```

death4	Odds Ratio	Std. Err.	z	P> z	[95% Conf. Interval]
bmi_middle	.2908277	.1409402	-2.55	0.011	.1124938 .7518705
bmi_high	.2154294	.1095045	-3.02	0.003	.0795486 .583415

```
. test bmi_middle+bmi_high=0

( 1) [death4]bmi_middle + [death4]bmi_high = 0

           chi2( 1) =      8.06
           Prob > chi2 =      0.0045
```

```
. test (bmi_middle=0) (bmi_high=0)

( 1) [death4]bmi_middle = 0
( 2) [death4]bmi_high = 0

           chi2( 2) =      9.63
           Prob > chi2 =      0.0081
```

Appendix for Question 3 (Appendix B)

```
. gen male_bmi = male*bmi
. gen male_age = male*age

. logit death4 age bmi male male_bmi male_age
```

```
Logistic regression                Number of obs   =      2500
                                   LR chi2(5)        =      121.78
                                   Prob > chi2        =      0.0000
Log likelihood = -725.12598        Pseudo R2      =      0.0775
```

death4	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
age	.1181702	.0179938	6.57	0.000	.082903	.1534374
bmi	.0474373	.0199491	2.38	0.017	.0083377	.0865368
male	7.397875	2.092839	3.53	0.000	3.295986	11.49976
male_bmi	-.0980824	.0324695	-3.02	0.003	-.1617214	-.0344434
male_age	-.0515358	.0231121	-2.23	0.026	-.0968347	-.0062369
_cons	-12.80381	1.564693	-8.18	0.000	-15.87056	-9.737071

```
. logistic death4 age bmi male male_bmi male_age
```

```
Logistic regression                Number of obs   =      2500
                                   LR chi2(5)        =      121.78
                                   Prob > chi2        =      0.0000
Log likelihood = -725.12598        Pseudo R2      =      0.0775
```

death4	Odds Ratio	Std. Err.	z	P> z	[95% Conf. Interval]	
age	1.125436	.0202509	6.57	0.000	1.086436	1.165835
bmi	1.04858	.0209183	2.38	0.017	1.008373	1.090392
male	1632.511	3416.584	3.53	0.000	27.00403	98692.45
male_bmi	.9065742	.029436	-3.02	0.003	.8506782	.9661431
male_age	.9497697	.0219512	-2.23	0.026	.9077061	.9937825