Homework 3 BSTA 513/613

Your name here - update this!!!!

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Directions

- Download the .qmd file here.
- You will need to download the datasets. Use this link to download the homework datasets needed in this assignment. If you do not want to make changes to the paths set in this document, then make sure the files are stored in a folder named "data" that is housed in the same location as this homework .qmd file.
- Please upload your homework to Sakai. Upload both your .qmd code file and the rendered .html file
 - Please rename you homework as Lastname_Firstinitial_HW0.qmd. This will help organize the homeworks when the TAs grade them.
 - Please also add the following line under subtitle: "BSTA 512/612": author: First-name Last-name with your first and last name so it is attached to the viewable document.
- For each question, make sure to include all code and resulting output in the html file to support your answers.
- Show the work of your calculations using R code within a code chunk. Make sure that both your code and output are visible in the rendered html file. This is the default setting.
- If you are computing something by hand, you may take a picture of your work and insert the image in this file. You may also use LaTeX to write it inline.
- Write all answers in complete sentences as if communicating the results to a collaborator. This means including a sentence summarizing results in the context of the research study.

💡 Tip

It is a good idea to try rendering your document from time to time as you go along! Note that rendering automatically saves your qmd file and rendering frequently helps you catch your errors more quickly.

Questions

Question 1

This question is taken from the Hosmer and Lemeshow textbook. The ICU study data set consists of a sample of 200 subjects who were part of a much larger study on survival of patients following admission to an adult intensive care unit (ICU). The dataset should be available within Course Materials. The major goal of this study was to develop a logistic regression model to predict the probability of survival to hospital discharge of these patients. In this question, the primary outcome variable is vital status at hospital discharge, STA. Clinicians associated with the study felt that a key determinant of survival was the patient's age at admission, AGE. We will be building to a multivariable logistic regression model while adjusting for cancer part of the present problem (CAN), CPR prior to ICU admission (CPR), infection probable at ICU admission (INF), and level of consciousness at ICU admission (LOC).

A code sheet for the variables to be considered is displayed in Table 1.5 below (from the Hosmer and Lemeshow textbook, pg. 23). We refer to this data set as the ICU data.

Variable	Description	Codes/Values	Name
1	Identification code	ID number	ID
2	Vital status at hospital discharge	0 = Lived	STA
		1 = Died	
3	Age	Years	AGE
4	Gender	0 = Male	GENDER
		1 = Female	
5	Race	1 = White	RACE
		2 = Black	
		3 = Other	
6	Service at ICU admission	0 = Medical	SER
		1 = Surgical	
7	Cancer part of present problem	0 = No	CAN
	F F	1 = Yes	
8	History of chronic renal failure	0 = No	CRN
	motory of emotion femal famale	1 = Yes	crut
9	Infection probable at ICU	0 = No	INF
	admission	1 = Yes	
10	CPR prior to ICU admission	0 = No	CPR
	er it prior to rece admission	1 = Yes	CIR
11	Systolic blood pressure at ICU	mm Hσ	SYS
	admission	IIIII IIg	515
12	Heart rate at ICU admission	Beats/min	ΗΡΔ
13	Previous admission to an ICU	0 - No	PRE
	within 6 months	1 - Yes	TRE
14	Type of admission	0 - Flective	TYPE
	Type of admission	1 - Emergency	THE
15	Long hone multiple neck single	0 - No	FRA
	area or hip fracture	0 = 100 1 - Ves	IKA
16	PO from initial blood gases	1 = 103 0 = >60	PO2
	102 nom mital blood gases	0 = > 00 1 = < 60	102
17	PH from initial blood gases	$1 = \le 00$ 0 = >7.25	DН
	FIT from mittal blood gases	$0 = \ge 7.25$ 1 = < 7.25	T II
18	PCO from initial blood gases	1 = < 7.23 0 = < 45	PCO
	100_2 from mittal blood gases	$0 = \leq 45$ 1 = ≤ 45	ico
19	Disorbonate from initial blood	1 = >43 0 - >18	DIC
		$0 = \ge 10$ 1 = <18	BIC
20	gases	1 = < 10	CDE
20	Creatinine from initial blood gases	$0 = \leq 2.0$	CKE
21	Level of consciousness at ICU	I = >2.0	LOC
	admission	$0 = No \ coma \ or$	LUC
	admission	1 Deer stupor	
		1 = Deep stupor	
		2 = Coma	

 Table 1.5
 Code Sheet for the Variables in the ICU Data

Part a

From the above list (AGE, CAN, CPR, INF, and LOC) of independent variables, identify if each is a continuous, binary, or multi-level (>2) categorical variable.

Part b

For the binary and multi-level categorical variables, please identify a reference group for each. Include justification for the reference group.

Part c

Refer back to Part c from Homework 2's Question 4. Interpret the odds ratio for age in the simple logistic regression model. Please include the 95% confidence interval.

Part d

Compute the predicted probability of hospital discharge for a subject who is 63 years old. Compute the 95% confidence interval for the predicted probability and interpret the predicted probability.

Part e

For the categorical variables (binary and multi-group), please mutate the variables within the ICU dataset to set your chosen reference groups.

Part f

Write down the equation for the logistic regression model of STA on CPR.

Part g

Using the glm() function, obtain the maximum likelihood estimates of the coefficient parameters of the logistic regression model in Part f. Using these estimates, write down fitted logistic regression model.

Part h

Write a sentence interpreting the odds ratio for the coefficients in Part g's model. Please include the 95% confidence interval.

Part i

Write down the equation for the logistic regression model of STA on LOC.

Part j

Using the glm() function, obtain the maximum likelihood estimates of the coefficient parameters of the logistic regression model in Part i. Present the coefficient estimates. No need to write out the fitted regression equation.

Please take note of the warnings that you receive from fitting the glm() model and any large coefficient estimate with large confidence intervals. In this case, we have a category within LOC that has very few observations. (We will discuss this more in Lesson 13: Numerical Problems)

Check the number of observations that have a deep stupor and death at discharge and the number of observations that have a deep stupor and live at discharge. You can do this using the table() function to create a contingency table.

Part k

Write a sentence interpreting the odds ratio of death for the indicator of coma. Please include the 95% confidence interval.