## Exam III

## STAT 430 FALL 2017

## Instructions

- The exam is a take home exam. It is due December 15th on blackboard by 5:00 pm.
- If you have any questions about, or need clarification on the meaning of an item on this exam, please ask your instructor during my office hours or email.
- No other form of external help is permitted attempting to receive help or provide help to others will be considered cheating.
- Do not cheat on this exam. Academic integrity demands an honest and fair testing environment. Cheating will not be tolerated and will result in an immediate score of 0 on the exam and an incident report will be submitted to the dean's office.

Name:			
Student ID:			

- 1. (5 points) Suppose that X is a Poisson random variable with rate  $\lambda$ . Find E(1/(X+1)).
- 2. (5 points) Suppose that X is a uniform on the interval [0, 1]. With  $Y = \sqrt{X}$ , find E(Y) and Var(Y).
- 3. Suppose that  $Z_1, Z_2, \ldots$  are independent standard normal random variables.
  - (a) (5 points) Find the moment generating function of  $X_n = \sum_{i=1}^n \frac{1}{3^i} Z_i$ .
  - (b) (5 points) Using the moment generating function, find the mean and variance of  $X_n$ .
  - (c) (5 points) As  $n \to \infty$ , what happens to the distribution of  $X_n$ ?
- 4. (10 points) In R, use the Monte Carlo method of integration to estimate the value of  $\int_0^1 \sin(2\pi x) dx$  with n = 100, n = 1000 and n = 10000. Compare the estimated integral to the exact value.
- 5. (5 points) (Edited Saturday, December 9 at 10:30)

Let  $\{X_i\}$  be a sequence of independent random variables with  $E(X_i) = \mu$  and  $Var(X_i) = \sigma_i^2$  (i.e., the variances of the random variables are different). Show that if  $\sum_{i=1}^n \sigma_i^2/n^2 \to 0$  then  $\bar{X}$  converges to  $\mu$  in probability.

note: when I refer to  $\bar{X}$  I could have used a better notation  $\bar{X}_n$  - it was intended to refer to  $\frac{1}{n}\sum_{i=1}^n X_i$ . As the we take new values in the sequence, we would produce a sequence of these "sample averages" - so  $\bar{X}_1, \bar{X}_2, \ldots$ 

6. (Edited Saturday, December 9 at 10:30)

Consider an sample of n independent random variables with density function

$$f(y|\theta) = \frac{\theta}{2}e^{-|y|/\theta}, -\infty < x < \infty$$

- (a) (5 points) Find the MME estimator of  $\theta$ .
- (b) (5 points) Find the MLE estimator of  $\theta$ .
- 7. Suppose that  $\epsilon_i$  are independent normal random variables with mean 0 and (unknown) variance  $\sigma^2$  for i = 1, 2, ..., n. For known values of  $x_i$ , consider the statistical model

$$y_i = \beta_0 + \beta_1 x_i + \beta_2 x_i^2 + \epsilon_i$$

- (a) (10 points) Find the maximum likelihood estimates  $\hat{\beta}_0$   $\hat{\beta}_1$   $\hat{\beta}_2$  and  $\hat{\sigma}^2$  in terms of the observable values of  $x_i$  and  $y_i$ .
- (b) (5 points) Suppose that an expiriment is performed and the following observations are collected: Using this data, provided the fitted version of the model from part a),

$$\hat{y}_i = \hat{\beta}_0 + \hat{\beta}_1 x_i + \hat{\beta}_2 x_i^2$$

- (c) (5 points) In R, create a plot with the fitted values vs the observed values (i.e., the fitted values  $\hat{y}_i$  on the x-axis). What does this plot indicate about the overall quality of the fitted model?
- (d) (5 points) Using the fitted model from part b), find the values of the residuals

$$e_i = y_i - \hat{y}_i$$

and provide a histogram of these values. What does the shape of the residuals histogram indicate about the assumptions we make when fitting the model we used in part b)?

8. Suppose that  $\epsilon_i$  are independent normal random variables with mean 0 and (unknown) variance  $\sigma^2$  for i = 1, 2, ..., n. For known values of  $x_i$ , consider two statistical models:

Model A

$$y_i = \beta_0 e^{\beta_1 x_i} + \epsilon_i$$
 
$$\epsilon_i \sim N(0, \sigma^2) \text{ (independent)}$$

Model B

$$log(y_i) = \alpha_0 + \alpha_1 x_i + \epsilon_i$$
  
 
$$\epsilon_i \sim N(0, \sigma^2) \text{ (independent)}$$

- (a) (10 points) Using maximum likelihood estimators, provided the estimated parameters of Model A in terms of  $x_i$  and  $y_i$ .
- (b) (10 points) Using maximum likelihood estimators, provided the estimated parameters of Model B in terms of  $x_i$  and  $y_i$ .
- (c) (5 points) Are these models equivalent? That is, can we model an exponential relationship between  $y_i$  and  $x_i$  by modeling a linear relationship between  $log(y_i)$  and  $x_i$ ? Explain.