

Exam 2 Name (please print) KEY

Math 483, Fall 2019 November 21, 2019

Show work and correct notation for full credit. Give numerical or simple fraction answers whenever possible.

Problem	1	2	3	4	5	6	Total
Earned							
Possible	10	10	10	10	10	10	50

(1.) a)  $\hat{\mu}_1 = \bar{X}, \hat{\mu}_2 = \bar{Y}$   
 $se(\hat{\mu}_1) = \frac{\sigma}{\sqrt{n}} = \frac{\bar{X}}{\sqrt{n}}, se(\hat{\mu}_2) = \frac{\bar{Y}}{\sqrt{n}}$   
 b)  $\bar{X} - \bar{Y} \pm z_{\alpha/2} \sqrt{\frac{(\bar{X})^2}{n} + \frac{(\bar{Y})^2}{n}}$   
 $= 0.6 \pm 1.96 \sqrt{\frac{5.5^2}{100} + \frac{4.9^2}{100}}$   
 $= 0.6 \pm 1.44$

Solve 5 problems out of 6. Cross one out.

1. Suppose  $X_1, \dots, X_n \sim \text{Exponential}(\mu_1), Y_1, \dots, Y_n \sim \text{Exponential}(\mu_2)$  and  $n = 100$ . The sample observations yielded  $\bar{X} = 5.5$  and  $\bar{Y} = 4.9$ .

- (a) Find point estimates  $\hat{\mu}_i$  and  $se(\hat{\mu}_i), i = 1, 2$ .
- (b) Find a 95% confidence interval for  $\mu_1 - \mu_2$ , based on Normal approximation.
- (c) Based on the above CI, do you believe that  $\mu_1 = \mu_2$ ?
- (d) Perform the Wald test for  $H_0: \mu_1 = \mu_2$ .

(c) yes,  
 $\mu_1 - \mu_2 = 0$   
 is in the C.I.

(d)  $W = \frac{0.6}{\sqrt{\frac{5.5^2}{100} + \frac{4.9^2}{100}}}$   
 $= 0.814$

$|W| < z_{\alpha/2} = 1.96$   
 $\Rightarrow$  accept  $H_0$   
 at  $\alpha = 0.05$ ;  $p\text{-value} = 2 P(Z > 0.814) = 2(0.5 - 0.2910) \approx .42$

2. MLE

Let  $X$  have the Binomial distribution with given  $n$  and unknown  $p = \theta$ .

- (a) Find the MLE for  $\theta$ . [Show the derivation!]
- (b) Find the Fisher information [Show the complete derivation!]
- (c) Describe the asymptotic distribution of the MLE as  $n$  gets large.

$L(\theta) = \binom{n}{x} \theta^x (1-\theta)^{n-x}$   
 $l(\theta) = \text{const} + x \ln \theta + (n-x) \ln(1-\theta)$   
 $l'(\theta) = \frac{x}{\theta} - \frac{n-x}{1-\theta} = 0$   
 $\Rightarrow x(1-\theta) = \theta(n-x)$   
 $\Rightarrow \hat{\theta} = \frac{x}{n} = \hat{p}$

b)  $I_n(\theta) = -E(l''(\theta)) = -E\left(-\frac{x}{\theta^2} - \frac{n-x}{(1-\theta)^2}\right) = \frac{n\theta}{\theta^2} + \frac{n(1-\theta)}{(1-\theta)^2}$   
 $= n\left(\frac{1}{\theta} + \frac{1}{1-\theta}\right) = \frac{n}{\theta(1-\theta)}$

c)  $\hat{\theta} \approx N\left(\theta, \frac{1}{I_n(\theta)}\right) = N\left(\theta, \frac{\theta(1-\theta)}{n}\right)$

3.  $\lambda = 2 \ln \left( \frac{L(\hat{\theta})}{L(\theta_0)} \right)$ , recall  $\hat{\theta} = \bar{X}$  for Normal.  
 $\ln L = \text{const} - \frac{\sum (X_i - \theta)^2}{2} \Rightarrow \lambda = \sum (X_i - 0)^2 - \sum (X_i - \bar{X})^2$

3. For the Normal distribution with unknown mean  $\mu = \theta$  and known  $\sigma = 1$ , perform the likelihood ratio test for

$$\begin{cases} H_0: \mu = 0 \\ H_1: \mu \neq 0 \end{cases}$$

based on the data:  $n = 50, \sum X_i = 14, \sum X_i^2 = 240$ .

Extra credit

Calculate the power of the Wald test for the above, with  $n = 50, \mu = -0.2$  and  $\alpha = 0.1$

$$\begin{aligned} &= \sum X_i^2 - \sum (X_i - \bar{X})^2 \\ &= \sum X_i^2 - n(\bar{X})^2 \\ &= \sum X_i^2 - \frac{(\sum X_i)^2}{n} \\ &= 240 - \frac{14^2}{50} \\ &= 238.44 \end{aligned}$$

Use  $\chi^2(1)$  table, at  $\alpha = 0.05$   
 $\chi^2_{1, \alpha} = 3.841$

$$238.44 > 3.841$$

$\Rightarrow$  Reject  $H_0$  at  $\alpha = 5\%$

E.C.  
 $\beta(\theta) = 1 - \Phi\left(\frac{\theta_0 - \theta_1}{\sigma} + z_{\alpha/2}\right) + \Phi\left(\frac{\theta_0 - \theta_1}{\sigma} - z_{\alpha/2}\right)$   
 $= 1 - \Phi(1.41 + 1.645) + \Phi(1.41 - 1.645)$   
 $\approx 0.41$

4. In a genetic study, the subjects were classified into 4 groups, with the counts as follows

Group	1	2	3	4	Total
Count	22	20	30	28	100
$E_i$	25	25	25	25	

Can the groups be equally distributed among the population? Test at the level  $\alpha = 0.05$ .

4.  
 $H_0: p_1 = p_2 = p_3 = p_4 = 1/4$

Table C: Critical points of the chi-square distribution

Degrees of freedom	Upper tail probability						
	0.100	0.050	0.025	0.010	0.005	0.001	0.0005
1	2.706	3.841	5.024	6.635	7.879	10.828	12.116
2	4.605	5.991	7.378	9.210	10.597	13.816	15.202
3	6.251	7.815	9.348	11.345	12.838	16.266	17.730
4	7.779	9.488	11.143	13.277	14.860	18.467	19.997
5	9.236	11.070	12.833	15.086	16.750	20.515	22.105

$$\begin{aligned} \chi^2 &= \sum \frac{(X_i - E_i)^2}{E_i} \\ &= \frac{(22-25)^2}{25} + \dots + \frac{(28-25)^2}{25} \\ &= 2.72 \end{aligned}$$

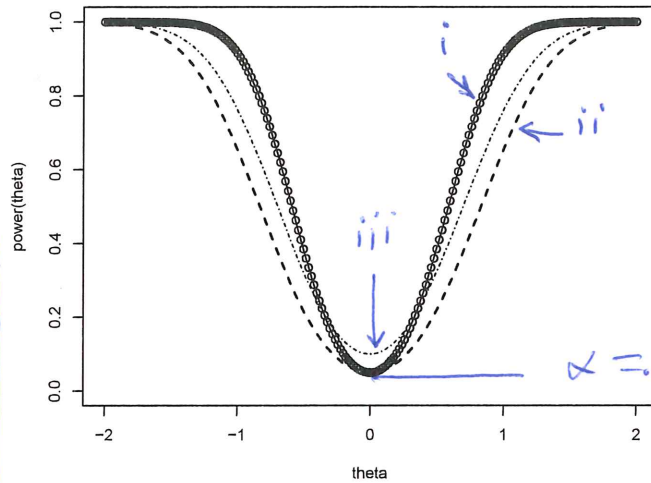
$df = k - 1 = 3$

at  $\alpha = 0.05, \chi^2_{0.05, 3} = 7.815 > 2.72$

$\Rightarrow$  Accept  $H_0$ , no evidence of difference.

5. (a) Match each of the power curves below with one of the following cases:

- i.  $n = 100, \alpha = 0.05$     ii.  $n = 50, \alpha = 0.05$     iii.  $n = 50, \alpha = 0.1$



$(\theta) g(\theta_1, \theta_2) = \theta_1, \theta_2$   
 $\hat{\theta}_1, \hat{\theta}_2 \approx \mathcal{N}(\theta_1, \theta_2, (\nabla g)^T J (\nabla g))$   
 $\nabla g = \begin{bmatrix} \theta_2 \\ \theta_1 \end{bmatrix} = \begin{bmatrix} 3 \\ 1 \end{bmatrix}$   
 $J = I^{-1} = \frac{1}{n} \begin{bmatrix} 1/2 & 0 \\ 0 & 2 \end{bmatrix}$   
 $\hat{\theta}_1, \hat{\theta}_2 \approx \mathcal{N}(3, \frac{13}{2n})$

(b)  $\hat{\theta} = [\hat{\theta}_1, \hat{\theta}_2]^T$  is an MLE for some estimation problem with the Fisher information matrix

$$I = n \begin{bmatrix} 2 & 0 \\ 0 & 0.5 \end{bmatrix}$$

Assuming that the "true values" are  $\theta_1 = 1$  and  $\theta_2 = 3$ , use Delta method to find the approximate distribution of  $\hat{\theta}_1, \hat{\theta}_2$

6. Misc.

(4 pt)

(a) Give an example of an estimation problem for which MLE is not unbiased.

$X_i \sim \mathcal{N}(\mu, \sigma^2)$   
 $\hat{\sigma}^2 = \frac{\sum (X_i - \bar{X})^2}{n}$   
 is biased  
 OR  
 $\hat{\theta} = \max\{X_1, \dots, X_n\}$   
 $E(\hat{\theta}) < \theta$

True or False? Explanations are not necessary, but they won't hurt.

(b) Fisher information matrix can be used to make a quadratic approximation to the log-likelihood surface.

T

(c) Since  $\hat{\sigma}^2$  is an MLE for  $\sigma^2$ , then  $\hat{\sigma}$  is a consistent estimate for  $\sigma$ .

T (MLE is consistent)

(d) The problem of finding the MLE for  $\theta$  when  $X_1, \dots, X_n \sim \text{Uniform}[0, \theta]$  is an example of a regular estimation problem.

F (Fisher info. DNE;  $\hat{\theta}$  not asymp. Normal)

2 pt each