

**Stat 543 Exam 3
Spring 2016**

I have neither given nor received unauthorized assistance on this exam.

KEY

Name Signed

Date

Name Printed

This exam consists of 11 questions. Do at least 9 of them. I will score answers at 10 points apiece and count you best 9 scores (making 90 points possible).

1. A particular continuous model for random pairs (X, Y) with parameter $\gamma \in (0, \infty)$ has joint pdf

$$f(x, y | \gamma) = \frac{\gamma}{2\pi\sqrt{x^2 + y^2}} \exp(-\gamma\sqrt{x^2 + y^2}) \quad \forall (x, y) \in \mathbb{R}^2$$

(This "radially symmetric" density is constant on circles centered at the origin.)

a) For n iid data pairs $(X_1, Y_1), \dots, (X_n, Y_n)$ identify a statistic in which there is monotone likelihood ratio (and show that your statistic really does the job).

Let $R_i = \sqrt{X_i^2 + Y_i^2}$. The likelihood is $\prod_{i=1}^n f(x_i, y_i | \gamma)$
 i.e. the likelihood is $\frac{\gamma^n}{(2\pi)^n \prod_{i=1}^n r_i} \exp(-\gamma(\sum r_i))$. So for

$\gamma_2 > \gamma_1$ the likelihood ratio is

$$\Lambda(\gamma_2, \gamma_1) = \left(\frac{\gamma_2}{\gamma_1}\right)^n \exp\left(\left[-\sum_{i=1}^n r_i\right](\gamma_2 - \gamma_1)\right)$$

which is monotone increasing in $(-\sum_{i=1}^n r_i)$

b) For the $n=1$ case, find the UMP size $\alpha = .05$ test of $H_0: \gamma \leq 1$ vs $H_a: \gamma > 1$ in as explicit a form as is possible. (You may recall that change of variables to polar coordinates in a double integral involves replacing $dx dy$ with $r dr d\theta$.)

We want to reject H_0 for large values of $T = -R$
 i.e. small values of R . The cut-off value c needs to be chosen so that $P_{\gamma=1}[R < c] = .05$

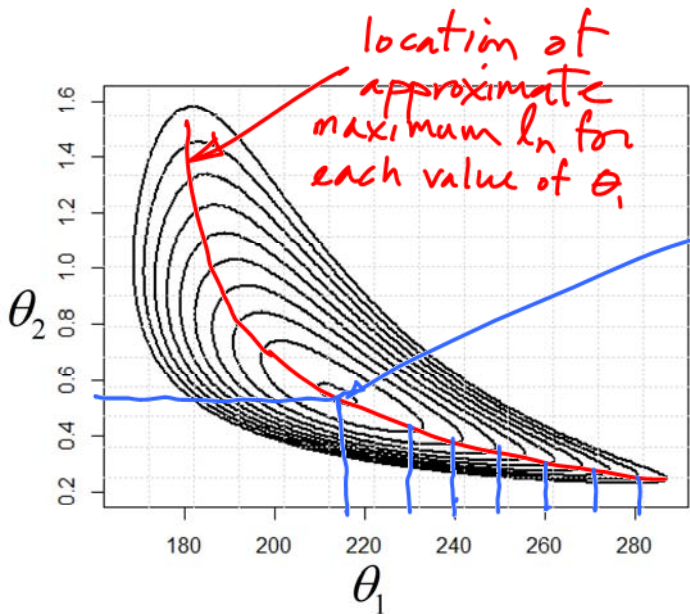
$$\begin{aligned} P_{\gamma=1}[R < c] &= \iint f(x, y | 1) dx dy \\ &= \int_0^{2\pi} \int_0^c \frac{1}{2\pi r} \exp(-r) r dr d\theta = 1 - e^{-c} \end{aligned}$$

So we want $.05 = 1 - e^{-c}$ i.e. $e^{-c} = .95$ i.e.

$c = -\ln(.95)$. The test is $\phi(x, y) = I[\sqrt{x^2 + y^2} < -\ln(.95)]$.

2. Below is a cartoon of a contour plot for a loglikelihood function $l_n(\theta_1, \theta_2)$. Find approximate values for a maximum likelihood estimate of $\theta = (\theta_1, \theta_2)$ and the values of the profile loglikelihood for θ_1 , $l_n^*(\theta_1)$. (The tight contours are at large values of the loglikelihood. Presume that the tightest contour is at

-10 and that they fall off at 5 units per contour, so that they are at -10, -15, -20, ...)



$\hat{\theta}^{MLE} \approx (215, .55)$

Contour loglikelihood values are:

$l_n^*(230) \approx -14.5$

$l_n^*(240) \approx -19.5$

$l_n^*(250) \approx -25.0$

$l_n^*(260) \approx -34$

$l_n^*(270) \approx -43$

$l_n^*(280) \approx -50$

3. A method of moments estimate of a parameter θ is $\hat{\theta} = 17$. The loglikelihood function, $l(\theta)$, is complicated, but it is possible to find (numerical) first and second derivatives $l'(17) = 1$ and $l''(17) = -.2$. Find a correction/improvement on the method of moments estimate based on these derivatives.



$l'(\theta) = 0$ has a root that can be approximated as

$17 + \Delta$ where $-\frac{1}{\Delta} = l''(17)$

i.e. $\Delta = \frac{1}{.2} = 5$

So the "improved" estimate is $17 + 5 = 22$.

4. A loglikelihood for a parameter vector $\theta = (\theta_1, \theta_2)$ is approximately quadratic near $\hat{\theta}^{\text{MLE}} = \left(\frac{1}{2}, \frac{1}{2}\right)$,

$$l_n(\theta_1, \theta_2) \approx -2\theta_1^2 + 2\theta_1\theta_2 - \theta_2^2 + \theta_1$$

Based on a large sample approximation to the distribution of a vector MLE, use the Wald method to give an approximately 95% two-sided confidence interval for θ_1 .

2nd partials of l_n are $\frac{\partial^2}{\partial \theta_1^2} l_n = \frac{\partial}{\partial \theta_1} (-4\theta_1 + 2\theta_2 + 1) = -4$
 $\frac{\partial^2}{\partial \theta_2^2} l_n = \frac{\partial}{\partial \theta_2} (2\theta_1 - 2\theta_2) = -2$ and $\frac{\partial^2}{\partial \theta_1 \partial \theta_2} l_n = 2$. So the inverse
 negative Hessian at $\left(\frac{1}{2}, \frac{1}{2}\right)$ is $\begin{pmatrix} 4 & -2 \\ -2 & 2 \end{pmatrix}^{-1} = \frac{1}{4} \begin{pmatrix} 2 & -2 \\ -2 & 4 \end{pmatrix} = \begin{pmatrix} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & 1 \end{pmatrix}$

So an estimated variance of the 1st co-ordinate of $\hat{\theta}^{\text{MLE}}$ is $\frac{1}{2}$. So an approximate 95% interval

is $\hat{\theta}_1^{\text{MLE}} \pm z \sqrt{\text{est'd variance}} \quad \text{i.e.} \quad \frac{1}{2} \pm 1.96 \sqrt{\frac{1}{2}}$

5. Suppose that $k(\lambda)$ is the smallest integer k so that for $X \sim \text{Poisson}(\lambda)$, $P[X \geq k] \leq .05$. You may think of plotting $k(\lambda)$ and producing an increasing step function taking integer values. If I observe a Poisson variable X , exactly how should I use that step function to produce a confidence interval for λ with associated confidence at least 95%?

We can invert a set of size $\alpha \leq .05$ tests of $H_0: \lambda = \lambda_0$. Test for λ_0 is $\mathbb{I}[X \geq k(\lambda_0)]$. Then the set of λ_0 for which H_0 is not rejected serves as a confidence set. This is the set of λ_0 for which $X < k(\lambda_0)$. This is the set of λ_0 for which the step function takes the value $X+1$ or larger.

6. For parameters $0 < c < 1$ and $0 < p < 1$, a pdf on $(0,1)$ is of the form

$$f(x|c,p) = \begin{cases} \frac{p}{c} & \text{for } 0 < x < c \\ \frac{1-p}{1-c} & \text{for } c < x < 1 \end{cases}$$

(The density is constant to the left and then to the right of c , assigning probabilities $p = P[X \leq c]$ and $1-p = P[X > c]$.) Find the form of the likelihood ratio statistic for testing $H_0 : c = .5$ based on n iid observations from this density, X_1, \dots, X_n .

For any c , let $n(c) = \# [x_i \leq c]$

$$\lambda(\underline{x}) = \frac{\max_{p,c} \prod_{i=1}^n f(x_i|c,p)}{\max_p \prod_{i=1}^n f(x_i|.5,p)}$$

For any given c , $\prod_{i=1}^n f(x_i|c,p) \propto p^{n(c)} (1-p)^{n-n(c)}$ which is maximized at $\hat{p}(c) \equiv \frac{n(c)}{n}$. So define

$$\hat{m}(c) = \left(\frac{\hat{p}(c)}{c} \right)^{n(c)} \left(\frac{1-\hat{p}(c)}{1-c} \right)^{n-n(c)}$$

The denominator of λ is $\hat{m}(\frac{1}{2})$. The numerator is $\max_c \hat{m}(c)$. Note that $n(c)$ and thus $n-n(c)$, $\hat{p}(c)$, and $1-\hat{p}(c)$ change only at X order statistics. For c between order statistics $c^{n(c)} (1-c)^{n-n(c)}$ is minimized at one of the order statistics. So $\hat{m}(c)$ is maximized by searching over $c = X_{(1)}, X_{(2)}, \dots, X_{(n)}$. This amounts to examining the n values.

$$\left(\frac{j/n}{X_{(j)}} \right)^j \left(\frac{1-j/n}{1-X_{(j)}} \right)^{n-j}$$

and choosing the j maximizing this, say j^* . Then

$$\lambda(\underline{x}) = \hat{m}\left(\frac{j^*}{n}\right) / \hat{m}\left(\frac{1}{2}\right)$$

7. A discrete distribution has pmf

$$f(x|p) = \begin{cases} \frac{1}{2}(2p - p^2) & x=1,2 \\ p(1-p)^{x-1} & x=3,4,\dots \end{cases}$$

Find a 2-dimensional sufficient statistic based on a sample X_1, X_2, \dots, X_n iid from this distribution, and say carefully how you know that it is sufficient.

Let $n^* = \# [x_i = 1, 2]$
 Then $\prod_{i=1}^n f(x_i|p) = \left(\frac{1}{2}(2p - p^2)\right)^{n^*} p^{n-n^*} (1-p)^{\sum_{x_i > 2} x_i - n^*}$
 which with $g(p, m, t) = \left(\frac{1}{2}(2p - p^2)\right)^m p^{n-m} (1-p)^{t-m}$
 is $g(p, n^*, \sum_{x_i > 2} x_i)$ so that by the factorization theorem
 theorem $(n^*, \sum_{x_i > 2} x_i)$ is sufficient for p

8. For a distribution on $\{1, 2, \dots, M\}$ specified by values p_1, p_2, \dots, p_M , the quantity

$$\mathcal{E}(\mathbf{p}) = -\sum_{m=1}^M p_m \ln(p_m)$$

is called the "entropy" of the distribution. Using a statistical "information" measure, show that this is no larger than $\ln(M)$ (the entropy of the uniform distribution). (In the above expression the convention that $0 \cdot \ln(0) = 0$ is in force.)

$$\begin{aligned} \mathcal{E}(\mathbf{p}) &= -\sum_{m=1}^M p_m \ln(p_m) - \sum_{m=1}^M p_m \ln(M) + \ln(M) \\ &= -\left(\sum_{m=1}^M p_m \frac{\ln(p_m)}{\ln(\frac{1}{M})}\right) + \ln(M) \\ &= -K(\mathbf{p}, \mathbf{u}) + \ln(M) \leq \ln(M) \end{aligned}$$

\nearrow
 K-L information is ≥ 0
 \uparrow
 uniform dsn on $\{1, 2, \dots, M\}$

9. A pmf on the integers with an integer parameter, θ , is

$$f(x|\theta) = \frac{1}{2} I[x = \theta - 1 \text{ or } x = \theta + 1]$$

For X_1 and X_2 iid from this distribution, compare MSE's for the two estimators of θ ,

$$\hat{\theta} = \begin{cases} X_1 + 1 & \text{if } X_1 = X_2 \\ \bar{X} & \text{if } X_1 \neq X_2 \end{cases} \quad \text{and} \quad \tilde{\theta} = \hat{\theta} - \frac{1}{2}$$

10. Suppose that $X \sim U\left(\theta - \frac{1}{2}, \theta + \frac{1}{2}\right)$ and that a prior distribution for θ is $N(0,1)$. Find the Bayes estimator of θ under squared error loss.