

BSc, MSci and MSc EXAMINATIONS (MATHEMATICS)

May-June 2016

This paper is also taken for the relevant examination for the Associateship of the
Royal College of Science

Statistical Theory 1

Date: Tuesday 10th May 2016

Time: 09.30 – 11.30

Time Allowed: 2 Hours

This paper has Four Questions.

Candidates should use ONE main answer book.

Supplementary books may only be used after the relevant main book(s) are full.

Statistical tables are provided.

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- Affix one of the labels provided to each answer book that you use, but DO NOT USE THE LABEL WITH YOUR NAME ON IT.
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Raw Mark	Up to 12	13	14	15	16	17	18	19	20
Extra Credit	0	$\frac{1}{2}$	1	$1\frac{1}{2}$	2	$2\frac{1}{2}$	3	$3\frac{1}{2}$	4

- Each question carries equal weight.
- Calculators may not be used.

- State the Neyman Factorization Criterion and prove it for the case of discrete distributions.
 - Suppose that $X_1, \dots, X_n \stackrel{\text{i.i.d.}}{\sim} \text{Bernoulli}(\theta)$, where $0 < \theta < 1$.
 - Show that $T = \sum_{i=1}^n X_i$ is a sufficient statistic for θ . Is T complete? Why?
Also, argue whether or not T is minimal sufficient for θ .
 - Find the UMVUE of $\theta(1 - \theta)$.
[Hint: $I(X_1 = 0, X_2 = 1)$ is an unbiased estimator of $\theta(1 - \theta)$.]
 - Compute the Cramér-Rao lower bound for the variance of unbiased estimators of $\theta(1 - \theta)$. Does the UMVUE of $\theta(1 - \theta)$ obtained in (ii) attain this lower bound? Why or why not?
- Let $(X_1, Y_1), \dots, (X_n, Y_n)$ be independent pairs of Normal random variables where X_i and Y_i are independent $N(\mu_i, \sigma^2)$ random variables.
 - Find the MLEs of μ_1, \dots, μ_n and σ^2 .
 - Now, suppose we observe only Z_1, \dots, Z_n where $Z_i = X_i - Y_i$.
 - Find the MLE of σ^2 based on Z_1, \dots, Z_n and discuss whether or not it is consistent.
 - Obtain a method of moments (MM) estimator of σ^2 based on Z_1, \dots, Z_n .
 - Consider testing $H_0 : \sigma^2 \leq \sigma_0^2$ versus $H_1 : \sigma^2 > \sigma_0^2$. Find the UMP test at level α based on Z_1, \dots, Z_n .
 - Is the UMP level α test obtained in (iii) unbiased? Justify your answer.

3. Let X_1, \dots, X_n be i.i.d. random variables from the delayed exponential distribution having the probability density function

$$f_\theta(x) = \theta e^{-\theta(x-2)}, \quad x > 2,$$

where θ is unknown. Suppose that the prior distribution for θ is $\text{Exponential}(\lambda)$ where λ is a known positive constant.

- (a) Obtain the posterior distribution of θ .
- (b) Is the prior here a conjugate prior? Justify your answer.
- (c) Find the Bayesian point estimator of θ under the squared error loss function.
- (d) Verify whether or not the Bayes estimator obtained in (c) is admissible.

4. Suppose that $X_1, \dots, X_m \stackrel{\text{i.i.d.}}{\sim} \text{Exponential}(\theta_1)$ and $Y_1, \dots, Y_n \stackrel{\text{i.i.d.}}{\sim} \text{Exponential}(\theta_2)$, and assume the X_i and the Y_i are independent. Consider testing $H_0 : \theta_1 = \theta_2$ versus $H_1 : \theta_1 \neq \theta_2$.

- (a) Show that the likelihood ratio test statistic is as follows

$$\lambda(x, y) = \left(\frac{m}{m+n} + \frac{n}{m+n} \frac{\bar{Y}}{\bar{X}} \right)^{-m} \left(\frac{n}{m+n} + \frac{m}{m+n} \frac{\bar{X}}{\bar{Y}} \right)^{-n}.$$

- (b) Obtain a test at level α using the test statistic $\frac{\bar{X}}{\bar{Y}}$.
[Hint: Use the fact that if χ_1^2 and χ_2^2 are two independent chi-squared random variables with degrees of freedom v_1 and v_2 respectively, then $\frac{\chi_1^2/v_1}{\chi_2^2/v_2} \sim F(v_1, v_2)$.]
- (c) Obtain the likelihood ratio test using the asymptotic distribution of $-2\log(\lambda(x, y))$ under H_0 .
- (d) Construct a confidence interval for $\frac{\theta_1}{\theta_2}$ with confidence coefficient $1 - \alpha$.

DISCRETE DISTRIBUTIONS						
	RANGE \mathbb{X}	PARAMETERS	MASS FUNCTION f_X	CDF F_X	$E_{f_X}[X]$	$\text{Var}_{f_X}[X]$
$Bernoulli(\theta)$	$\{0,1\}$	$\theta \in (0,1)$	$\theta^x(1-\theta)^{1-x}$		θ	$\theta(1-\theta) + \theta e^\theta$
$Binomial(n,\theta)$	$\{0,1,..,n\}$	$n \in \mathbb{Z}^+, \theta \in (0,1)$	$\binom{n}{x} \theta^x(1-\theta)^{n-x}$		$n\theta$	$(1 - \theta + \theta e^\theta)^n$
$Poisson(\lambda)$	$\{0,1,2,..\}$	$\lambda \in \mathbb{R}^+$	$\frac{e^{-\lambda} \lambda^x}{x!}$		λ	$\exp\{\lambda(e^\lambda - 1)\}$
$Geometric(\theta)$	$\{1,2,..\}$	$\theta \in (0,1)$	$(1-\theta)^{x-1}\theta$	$1 - (1-\theta)^x$	$\frac{1}{\theta}$	$\frac{(1-\theta)}{\theta^2} + \frac{\theta e^\theta}{1-e^\theta(1-\theta)}$
$NegBinomial(n,\theta)$ or	$\{n,n+1,..\}$ $\{0,1,2,..\}$	$n \in \mathbb{Z}^+, \theta \in (0,1)$ $n \in \mathbb{Z}^+, \theta \in (0,1)$	$\binom{x-1}{n-1} \theta^n (1-\theta)^{x-n}$ $\binom{n+x-1}{x} \theta^n (1-\theta)^x$		$\frac{n}{\theta}$ $\frac{n(1-\theta)}{\theta^2}$	$\left(\frac{\theta e^\theta}{1-e^\theta(1-\theta)}\right)^n$ $\left(\frac{\theta}{1-e^\theta(1-\theta)}\right)^n$

For CONTINUOUS distributions (see over), define the GAMMA FUNCTION

$$\Gamma(\alpha) = \int_0^\infty x^{\alpha-1} e^{-x} dx$$

and the LOCATION/SCALE transformation $Y = \mu + \sigma X$ gives

$$f_Y(y) = f_X\left(\frac{y-\mu}{\sigma}\right) \frac{1}{\sigma} \quad F_Y(y) = F_X\left(\frac{y-\mu}{\sigma}\right) \quad M_Y(t) = e^{\mu t} M_X(\sigma t) \quad E_{f_Y}[Y] = \mu + \sigma E_{f_X}[X] \quad \text{Var}_{f_Y}[Y] = \sigma^2 \text{Var}_{f_X}[X]$$

CONTINUOUS DISTRIBUTIONS						
	PARAMS.	PDF	CDF	$E_{f_X}[X]$	$\text{Var}_{f_X}[X]$	MGF
$Uniform(\alpha, \beta)$ (standard model $\alpha = 0, \beta = 1$)	\mathbb{X} (α, β)	$\frac{1}{\beta - \alpha}$	F_X $\frac{x - \alpha}{\beta - \alpha}$	$\frac{(\alpha + \beta)}{2}$	$\frac{(\beta - \alpha)^2}{12}$	M_X $\frac{e^{\lambda t} - e^{\alpha t}}{t(\beta - \alpha)}$
$Exponential(\lambda)$ (standard model $\lambda = 1$)	\mathbb{R}^+	$\lambda \in \mathbb{R}^+$	$\lambda e^{-\lambda x}$	$1 - e^{-\lambda x}$	$\frac{1}{\lambda}$	$\left(\frac{\lambda}{\lambda - t}\right)^n$
$Gamma(\alpha, \beta)$ (standard model $\beta = 1$)	\mathbb{R}^+	$\alpha, \beta \in \mathbb{R}^+$	$\frac{\beta^\alpha}{\Gamma(\alpha)} x^{\alpha-1} e^{-\beta x}$		$\frac{\alpha}{\beta^2}$	$\left(\frac{\beta}{\beta - t}\right)^n$
$Weibull(\alpha, \beta)$ (standard model $\beta = 1$)	\mathbb{R}^+	$\alpha, \beta \in \mathbb{R}^+$	$\alpha \beta x^{\alpha-1} e^{-\beta x^\alpha}$	$1 - e^{-\beta x^\alpha}$	$\frac{\Gamma(1 + 1/\alpha)}{\beta^{1/\alpha}}$	$\frac{\Gamma(1 + 2/\alpha) - \Gamma(1 + 1/\alpha)^2}{\beta^2 f_\alpha}$
$Normal(\mu, \sigma^2)$ (standard model $\mu = 0, \sigma = 1$)	\mathbb{R}	$\mu \in \mathbb{R}, \sigma \in \mathbb{R}^+$	$\frac{1}{\sqrt{2\pi\sigma^2}} \exp\left\{-\frac{(x - \mu)^2}{2\sigma^2}\right\}$		μ	$e^{\{\mu t + \sigma^2 t^2/2\}}$
$Student(\nu)$	\mathbb{R}	$\nu \in \mathbb{R}^+$	$\frac{(\pi\nu)^{-\frac{1}{2}} \Gamma\left(\frac{\nu+1}{2}\right)}{\Gamma\left(\frac{\nu}{2}\right) \left\{1 + \frac{x^2}{\nu}\right\}^{(\nu+1)/2}}$		0 (if $\nu > 1$)	$\frac{\nu}{\nu - 2}$ (if $\nu > 2$)
$Pareto(\theta, \alpha)$	\mathbb{R}^+	$\theta, \alpha \in \mathbb{R}^+$	$\frac{\alpha\theta^\alpha}{(\theta + x)^{\alpha+1}}$	$1 - \left(\frac{\theta}{\theta + x}\right)^\alpha$	$\frac{\theta}{\alpha - 1}$ (if $\alpha > 1$)	$\frac{\alpha\theta^2}{(\alpha - 1)(\alpha - 2)}$ (if $\alpha > 2$)
$Beta(\alpha, \beta)$	$(0, 1)$	$\alpha, \beta \in \mathbb{R}^+$	$\frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha)\Gamma(\beta)} x^{\alpha-1} (1 - x)^{\beta-1}$		$\frac{\alpha}{\alpha + \beta}$	$(\alpha + \beta)^2 (\alpha + \beta + 1)$

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- Each question carries equal weight.
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- State the Neyman Factorization Criterion and prove it for the case of discrete distributions.
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where θ is unknown. Suppose that the prior distribution for θ is $\text{Exponential}(\lambda)$ where λ is a known positive constant.

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- (c) Obtain the likelihood ratio test using the asymptotic distribution of $-2\log(\lambda(x, y))$ under H_0 .
- (d) Construct a confidence interval for $\frac{\theta_1}{\theta_2}$ with confidence coefficient $1 - \alpha$.

Mastery Question:

5. Let X_1, \dots, X_n be i.i.d. Cauchy random variables with density function

$$f_\theta(x) = \frac{1}{\pi (1 + (x - \theta)^2)} \quad x \in R,$$

and suppose outcomes x_1, \dots, x_n are observed.

- (a) Write down the likelihood equation for estimating θ and discuss whether it has a unique solution for the given sample x_1, \dots, x_n .
- (b) Given an estimate $\hat{\theta}^{(k)}$ for θ at iteration k , obtain a new estimate $\hat{\theta}^{(k+1)}$ using the Newton-Raphson method.
- (c) Show that a new estimate $\hat{\theta}^{(k+1)}$ using the Fisher scoring algorithm is

$$\hat{\theta}^{(k+1)} = \hat{\theta}^{(k)} + \frac{4}{n} \sum_{i=1}^n \frac{x_i - \hat{\theta}^{(k)}}{1 + (x_i - \hat{\theta}^{(k)})^2}.$$

[Hint: $\int_0^\infty \frac{1-t^2}{(1+t^2)^3} dt = \frac{\pi}{8}$.]

- (d) Is the sample mean \bar{x} an appropriate initial value for the Newton-Raphson and the Fisher scoring methods here? If not, suggest a good starting point. Briefly explain your thinking.
- (e) Which method has a faster convergence: the Newton-Raphson method or the Fisher scoring algorithm? Why?

DISCRETE DISTRIBUTIONS

	RANGE \mathbb{X}	PARAMETERS	MASS FUNCTION f_X	CDF F_X	$\mathbb{E}_{f_X}[X]$	$\text{Var}_{f_X}[X]$	MGF M_X
<i>Bernoulli(θ)</i>	$\{0, 1\}$	$\theta \in [0, 1]$	$\theta^x(1-\theta)^{1-x}$		θ	$\theta(1-\theta)$	$1 - \theta + \theta e^t$
<i>Binomial(n, θ)</i>	$\{0, 1, \dots, n\}$	$n \in \mathbb{Z}^+, \theta \in (0, 1)$	$\binom{n}{x} \theta^x(1-\theta)^{n-x}$		$n\theta$	$n\theta(1-\theta)$	$(1 - \theta + \theta e^t)^n$
<i>Poisson(λ)</i>	$\{0, 1, 2, \dots\}$	$\lambda \in \mathbb{R}^+$	$\frac{e^{-\lambda}\lambda^x}{x!}$		λ	λ	$\exp\{\lambda(e^t - 1)\}$
<i>Geometric(θ)</i>	$\{1, 2, \dots\}$	$\theta \in (0, 1)$	$(1-\theta)^{x-1}\theta$	$1 - (1-\theta)^x$	$\frac{1}{\theta}$	$\frac{(1-\theta)}{\theta^2}$	$\frac{\theta e^t}{1 - e^t(1-\theta)}$
<i>NegBinomial(n, θ)</i> or	$\{n, n+1, \dots\}$ $\{0, 1, 2, \dots\}$	$n \in \mathbb{Z}^+, \theta \in (0, 1)$	$\binom{x-1}{n-1} \theta^n(1-\theta)^{x-n}$ $\binom{n+x-1}{x} \theta^n(1-\theta)^x$		$\frac{n}{\theta}$ $\frac{n(1-\theta)}{\theta^2}$	$\frac{n(1-\theta)}{\theta^2}$ $\frac{n(1-\theta)}{\theta^2}$	$\left(\frac{\theta e^t}{1 - e^t(1-\theta)}\right)^n$ $\left(\frac{\theta}{1 - e^t(1-\theta)}\right)^n$

For CONTINUOUS distributions (see over), define the GAMMA FUNCTION

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and the LOCATION/SCALE transformation $Y = \mu + \sigma X$ gives

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CONTINUOUS DISTRIBUTIONS							
	PARAMS.	PDF	CDF	$\mathbb{E}_{f_X}[X]$	$\text{Var}_{f_X}[X]$	MGF	
\mathbb{X}		f_X	F_X				
$Uniform(\alpha, \beta)$ (standard model $\alpha = 0, \beta = 1$)	$\alpha < \beta \in \mathbb{R}$	$\frac{1}{\beta - \alpha}$	$\frac{x - \alpha}{\beta - \alpha}$	$\frac{(\theta + \beta)^2}{12}$	$\frac{e^{\theta t} - e^{\alpha t}}{t(\beta - \alpha)}$		
$Exponential(\lambda)$ (standard model $\lambda = 1$)	$\lambda \in \mathbb{R}^+$	$\lambda e^{-\lambda x}$	$1 - e^{-\lambda x}$	$\frac{1}{\lambda}$	$\frac{1}{\lambda^2}$	$\left(\frac{\lambda}{\lambda - t}\right)^\alpha$	
$Gamma(\alpha, \beta)$ (standard model $\beta = 1$)	$\alpha, \beta \in \mathbb{R}^+$	$\frac{\beta^\alpha}{\Gamma(\alpha)} x^{\alpha-1} e^{-\beta x}$		$\frac{\alpha}{\beta}$	$\frac{\sigma}{\beta^2}$	$\left(\frac{\beta}{\beta - t}\right)^\alpha$	
$Weibull(\alpha, \beta)$ (standard model $\beta = 1$)	$\alpha, \beta \in \mathbb{R}^+$	$\alpha \beta x^{\alpha-1} e^{-\beta x^\alpha}$	$1 - e^{-\beta x^\alpha}$	$\frac{\Gamma(1+1/\alpha)}{\beta^{1/\alpha}}$	$\frac{\Gamma(1+2/\alpha) - \Gamma(1+1/\alpha)^2}{\beta^{2/\alpha}}$		
$Normal(\mu, \sigma^2)$ (standard model $\mu = 0, \sigma = 1$)	$\mu \in \mathbb{R}, \sigma \in \mathbb{R}^+$	$\frac{1}{\sqrt{2\pi\sigma^2}} \exp\left\{-\frac{(x - \mu)^2}{2\sigma^2}\right\}$		μ	σ^2	$e^{\{\mu t + \sigma \sqrt{t}/2\}}$	
$Student(\nu)$	$\nu \in \mathbb{R}^+$		$\frac{(\pi\nu)^{-\frac{1}{2}} \Gamma\left(\frac{\nu+1}{2}\right)}{\Gamma\left(\frac{\nu}{2}\right) \left\{1 + \frac{x^2}{\nu}\right\}^{(\nu+1)/2}}$		$0 \quad (\text{if } \nu > 1)$ $\frac{\nu}{\nu-2} \quad (\text{if } \nu > 2)$		
$Pareto(\theta, \alpha)$	$\theta, \alpha \in \mathbb{R}^+$	$\frac{\alpha \theta^\alpha}{(\theta + x)^{\alpha+1}}$	$1 - \left(\frac{\theta}{\theta + x}\right)^\alpha$	$\frac{\theta}{\alpha-1}$ $\frac{(\alpha-1)(\sigma-2)}{(\text{if } \alpha > 2)}$ $\{\text{if } \alpha > 1\}$			
$Beta(\alpha, \beta)$	$(0, 1)$	$\alpha, \beta \in \mathbb{R}^+$	$\frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} x^{\alpha-1} (1-x)^{\beta-1}$	$\frac{\alpha\beta}{\alpha+\beta}$	$\frac{\alpha\beta}{(\alpha+\beta)^2(\alpha+\beta+1)}$		