

McGill University

Faculty of Science

MATH 556

MATHEMATICAL STATISTICS I

Final Examination

Date: 4th December 2008      Time: 9am-12pm

Examiner: Prof. D. A. Stephens      Associate Examiner: Prof. R. J. Steele

This paper contains six questions.

Credit will be given for all questions attempted.

Calculators may not be used. A Formula Sheet is provided.

1. (a) Suppose that  $X_1$  and  $X_2$  are independent random variables with  $Gamma(\alpha_1, 1)$  and  $Gamma(\alpha_2, 1)$  distributions respectively, for parameters  $\alpha_1, \alpha_2 > 0$ . Let

$$Y_1 = \frac{1}{X_1} \quad Y_2 = \frac{1}{X_2}$$

- (i) Find the pdf of  $Y_1$ .

4 MARKS

- (ii) Find the expectation and variance of  $Y_2$ . State explicitly conditions for the expectation and variance to exist.

6 MARKS

The distribution of  $Y_1$  and  $Y_2$  is termed the *inverse-Gamma* distribution.

- (b) Find the joint pdf of random variables

$$Z_1 = X_1 + X_2 \quad Z_2 = Y_1 + Y_2$$

and show that marginally  $Z_1$  has a Gamma distribution.

5 MARKS

Marginally, does  $Z_2$  have an inverse-Gamma distribution? Justify your answer.

5 MARKS

2. (a) For a scalar random variable  $X$ , the *cumulant generating function* (cgf),  $K_X$ , is defined in terms of the moment generating function,  $M_X$ , by

$$K_X(t) = \log M_X(t) \quad t \in (-h, h), \text{ some } h > 0$$

Find expressions for the expectation and variance of  $X$  in terms of  $K_X$ .

6 MARKS

- (b) For random variable  $X$ , consider the one parameter Exponential Family distribution,

$$f_X(x|\eta) = h(x)c(\eta) \exp\{\eta x\}$$

Find the form of  $K_X(t)$ , and an expression for  $\mathbb{E}_{f_X}[X]$  in terms of one or more of the functions and parameters that appear in the pdf.

8 MARKS

- (c) Show how the cumulant generating function plays a role in the *exponential tilting construction* of the Exponential Family of distributions. Illustrate your description with specific reference to the  $Normal(\theta, 1)$  distribution.

6 MARKS

3. (a) Consider scalar random variable  $X$ .

(i) Define the *characteristic function* of  $X$ ,  $C_X(t)$ .

2 MARKS

(ii) Show that  $|C_X(t)| \leq 1$  for all  $t \in \mathbb{R}$ .

2 MARKS

(iii) Describe how to diagnose whether a characteristic function corresponds to a discrete or a continuous distribution.

4 MARKS

(iv) State the *inversion formula* for a characteristic function known to belong to a **discrete** distribution.

2 MARKS

(b) Find  $C_X(t)$  if  $X$  is a continuous random variable with

$$f_X(x) = \exp\{-x - e^{-x}\} \quad x \in \mathbb{R}.$$

5 MARKS

(c) Find  $f_X(x)$  if  $C_X(t)$  is given by

$$C_X(t) = 1 - |t| \quad -1 < t < 1$$

and zero otherwise.

5 MARKS

4. (a) State and prove Jensen's Inequality in the univariate case. Define the Kullback-Leibler divergence between two pdfs  $f_1$  and  $f_2$  each with support  $\mathbb{R}$ ,  $\mathbb{K}(f_1, f_2)$ , and prove that

$$\mathbb{K}(f_1, f_2) \geq 0.$$

10 MARKS

(b) Suppose that  $X \sim Normal(0, \sigma^2)$ . Find a function of  $\sigma$ ,  $l(\sigma)$ , such that

$$P_X[-2 \leq X \leq 2] \geq l(\sigma)$$

which does **not** involve the standard normal cdf. Justify your answer.

4 MARKS

(c) If  $X \sim Poisson(\mu)$ , show that

$$P_X[X \geq 2\mu] \leq e^{\mu(e-3)}$$

Justify your answer.

6 MARKS

5. (a) A *finite mixture model* is a specific form of hierarchical model whose density,  $f_Y$ , takes the form

$$f_Y(y|\mathcal{T}, \boldsymbol{\theta}, L) = \sum_{l=1}^L f_l(y|\theta_l)\pi_l.$$

Explain the components of this specification, that is, the functions  $f_1, \dots, f_L$ , the parameters  $\theta_1, \dots, \theta_L$ , and the parameters  $\pi_1, \dots, \pi_L$ .

4 MARKS

Find the form of the moment generating function,  $M_Y$ , corresponding to  $f_Y$ .

4 MARKS

- (b) Suppose that  $X_1, \dots, X_n$  is a random sample from a  $Normal(\mu, 1)$  distribution. Find the distribution of the statistic

$$V = \sum_{i=1}^n (X_i - \bar{X})^2$$

You may quote without proof results concerning the sampling distribution of statistics derived from a normal random sample.

8 MARKS

- (c) Suppose that  $X_1, \dots, X_n$  is a random sample from an  $Exponential(\lambda)$  distribution. Find the distribution of the statistic

$$Y_1 = \min\{X_1, \dots, X_n\}$$

You may quote without proof results concerning the sampling distribution of order statistics derived from a random sample.

4 MARKS

6. (a) Suppose  $\{X_n\}$  are an independent sequence of random variables with cdf

$$F_X(x) = \frac{1}{1 + e^{-x}} \quad x \in \mathbb{R}$$

Let  $Y_n = \max\{X_1, \dots, X_n\}$ . Show that for large  $n$  and  $y > 0$ ,

$$P[Y_n > y] \simeq 1 - \exp\{-ne^{-y}\}.$$

6 MARKS

- (b) Suppose  $\{X_n\}$  are an independent sequence of  $Poisson(\lambda)$  random variables. Let

$$M_n = \frac{1}{n} \sum_{i=1}^n X_i$$

Show that  $M_n \xrightarrow{p} \lambda$  as  $n \rightarrow \infty$ . If random variable  $T_n$  is defined by  $T_n = e^{-M_n}$ , find an approximation to the distribution of  $T_n$  for large  $n$ .

8 MARKS

- (c) Suppose that  $X \sim Gamma(\alpha, 1)$ . By considering the variable

$$Z_\alpha = \frac{X - \alpha}{\sqrt{\alpha}}$$

or otherwise, construct an approximation to the distribution of  $Z_\alpha$  for large  $\alpha$ .

6 MARKS

**DISCRETE DISTRIBUTIONS**

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

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

$$f_Y(y) = f_X\left(\frac{y - \mu}{\sigma}\right) \frac{1}{\sigma} \qquad F_Y(y) = F_X\left(\frac{y - \mu}{\sigma}\right)$$

$$M_Y(t) = e^{\mu t} M_X(\sigma t)$$

$$E_{f_Y} [Y] = \mu + \sigma E_{f_X} [X]$$

$$\text{Var}_{f_Y} [Y] = \sigma^2 \text{Var}_{f_X} [X]$$

**CONTINUOUS DISTRIBUTIONS**

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