

McGill University

Course: MATH 556
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Faculty of Science

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MATH 556

MATHEMATICAL STATISTICS I

Setter	Checker	Associate Examiner
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MATHEMATICAL STATISTICS I

Final Examination

Date: 8th December 2006

Time: 2pm-5pm

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 random variable X has a *Gamma* $(\alpha, 1)$ distribution, for parameter $\alpha > 0$. Find the probability density function (pdf) of random variable Y defined by

$$Y = \frac{1}{X}.$$

Find also the expectation of Y .

- (b) Suppose that continuous random variable U has a cumulative distribution function (cdf), F_U , given by

$$F_U(u) = \exp\{-\exp\{-u\}\} \quad u \in \mathbb{R}.$$

Find the pdf of random variable V defined by

$$V = U^2.$$

- (c) Suppose that X and Y are positive, independent continuous random variables with cdfs F_X and F_Y and pdfs f_X and f_Y respectively. Show that

$$P[X < Y] = \iint_A f_X(x) f_Y(y) dx dy$$

for a suitably defined set A .

By considering the ranges of integration carefully, and making a change of variables, deduce that

$$P[X < Y] = \int_0^1 F_X(F_Y^{-1}(t)) dt.$$

where F_Y^{-1} is the inverse function for the 1-1 function F_Y .

2. (a) Suppose that Z_1 and Z_2 are independent random variables each having an *Exponential*(1) distribution. Find the joint pdf of random variables Y_1 and Y_2 defined by

$$Y_1 = \frac{Z_1}{Z_1 + Z_2} \quad Y_2 = Z_1 + Z_2.$$

- (b) The joint pmf/pdf of random variables X and Y can be specified in the following way:

$$f_{X,Y}(x, y) = f_{X|Y}(x|y) f_Y(y).$$

Find the marginal pmf of X

- (i) if

$$X|Y = y \sim \text{Binomial}(n, y)$$

for positive integer n , and continuous random variable Y has a standard uniform distribution, $Y \sim \text{Uniform}(0, 1)$;

- (ii) if

$$X|Y = y \sim \text{Exponential}(y)$$

and $Y \sim \text{Exponential}(\beta)$ for parameter $\beta > 0$.

3. (a) For a scalar random variable X , define the cumulant generating function (cgf), K_X , and show that

$$K_X^{(1)}(0) = E_{f_X}[X] \quad K_X^{(2)}(0) = Var_{f_X}[X]$$

where $K_X^{(r)}(t)$ denotes the r th derivative of K_X with respect to t .

Assume that K_X exists, and quote without proof properties of the moment generating function (mgf), M_X .

- (b) Suppose that continuous random variable X has pdf given by

$$f_X(x) = c \exp\{-\lambda|x|\} \quad x \in \mathbb{R}$$

for parameter $\lambda > 0$, and constant c . Find the characteristic function (cf) for X , $C_X(t)$.

- (c) Prove that the standard Normal distribution is infinitely divisible.

- (d) Suppose that random variable Y has cf defined by

$$C_Y(t) = \cos(t) \quad t \in \mathbb{R}.$$

Find the skewness of Y , ς , where

$$\varsigma = \frac{E_{f_Y}[(Y - \mu)^3]}{\sigma^3}$$

where μ and σ^2 are the expectation and variance of Y .

4. (a) Suppose that real constants a, b, p, q satisfy $a, b > 0$ and $p, q > 1$ with

$$\frac{1}{p} + \frac{1}{q} = 1.$$

Prove that

$$\frac{1}{p} a^p + \frac{1}{q} b^q \geq ab$$

with equality if and only if $a^p = b^q$.

Hence prove Hölder's Inequality: If X and Y are two random variables, and $p, q > 1$ satisfy the above identity,

$$|E_{f_{X,Y}}[XY]| \leq E_{f_{X,Y}}[|XY|] \leq \{E_{f_X}[|X|^p]\}^{1/p} \{E_{f_Y}[|Y|^q]\}^{1/q}$$

- (b) Prove that for random variables X and Y having joint pdf $f_{X,Y}$,

$$\{Cov_{f_{X,Y}}[X, Y]\}^2 \leq Var_{f_X}[X] Var_{f_Y}[Y]$$

5. (a) Show that the $Poisson(\lambda)$ and $Binomial(n, \theta)$ distributions are one parameter Exponential Family Distributions, and in each case find the natural (or canonical) parameter.
- (b) Consider the following three level hierarchical model

LEVEL 3: $r \in \{1, 2, \dots\}$ Fixed parameter

LEVEL 2: $V \sim Gamma(r/2, r/2)$

LEVEL 1: $X|V = v \sim Normal(0, v^{-1})$

Find the kurtosis of X , κ , defined by

$$\kappa = \frac{E_{f_X}[(X - \mu)^4]}{\sigma^4}$$

where μ and σ^2 are the expectation and variance of X . State precisely conditions on r for the kurtosis to be finite.

6. (a) Suppose that continuous random variables X_1, \dots, X_n are independent and identically distributed with cdf F_X specified by

$$F_X(x) = \frac{x^2}{1+x^2} \quad x > 0.$$

and zero otherwise. Let Y_n be the maximum order statistic derived from X_1, \dots, X_n , that is,

$$Y_n = \max\{X_1, \dots, X_n\}$$

Show that, in the limit as $n \rightarrow \infty$, the limiting distribution of Y_n does not exist, but that the limiting distribution of Z_n defined by

$$Z_n = Y_n/\sqrt{n}$$

does exist and is a continuous distribution.

- (b) Suppose that random variable Y_n can be written as

$$Y_n = \sum_{i=0}^n X_i$$

where, for each $i > 0$

$$X_i = \begin{cases} a & \text{with probability } \frac{1}{2} \\ -a & \text{with probability } \frac{1}{2} \end{cases}$$

and for $i = 0$, $X_0 = x_0$, a known constant.

- (i) Find an approximation to the distribution of Y_n for large n .
- (ii) Show that, for any n ,

$$P[|Y_n - x_0| \geq 2\sigma_n] \leq \frac{1}{4}$$

for some σ_n to be defined.

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> (θ)	$\{0, 1\}$	$\theta \in (0, 1)$	$\theta^x(1 - \theta)^{1-x}$		θ	$\theta(1 - \theta)$	$1 - \theta + \theta e^t$
<i>Binomial</i> (n, θ)	$\{0, 1, \dots, n\}$	$n \in \mathbb{Z}^+, \theta \in (0, 1)$	$\binom{n}{x} \theta^x (1 - \theta)^{n-x}$		$n\theta$	$n\theta(1 - \theta)$	$(1 - \theta + \theta e^t)^n$
<i>Poisson</i> (λ)	$\{0, 1, 2, \dots\}$	$\lambda \in \mathbb{R}^+$	$\frac{e^{-\lambda} \lambda^x}{x!}$		λ	λ	$\exp \{ \lambda (e^t - 1) \}$
<i>Geometric</i> (θ)	$\{1, 2, \dots\}$	$\theta \in (0, 1)$	$(1 - \theta)^{x-1} \theta$	$1 - (1 - \theta)^x$	$\frac{1}{\theta}$	$\frac{(1 - \theta)}{\theta^2}$	$\frac{\theta e^t}{1 - e^t(1 - \theta)}$
<i>NegBinomial</i> (n, θ)	$\{n, n + 1, \dots\}$	$n \in \mathbb{Z}^+, \theta \in (0, 1)$	$\binom{x-1}{n-1} \theta^n (1 - \theta)^{x-n}$		$\frac{n}{\theta}$	$\frac{n(1 - \theta)}{\theta^2}$	$\left(\frac{\theta e^t}{1 - e^t(1 - \theta)} \right)^n$
or	$\{0, 1, 2, \dots\}$	$n \in \mathbb{Z}^+, \theta \in (0, 1)$	$\binom{n+x-1}{x} \theta^n (1 - \theta)^{x-n}$		$\frac{n(1 - \theta)}{\theta}$	$\frac{n(1 - \theta)}{\theta^2}$	$\left(\frac{\theta}{1 - e^t(1 - \theta)} \right)^n$

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

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

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

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

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

CONTINUOUS DISTRIBUTIONS

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