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ID:

McGill University

Faculty of Science

Final Examination

MATH 556: Mathematical Statistics I

Examiner: Professor J. Nešlehová

Date: Friday, December 9, 2010

Associate Examiner: Professor D. A. Stephens

Time: 9:00 A.M. – 12:00 P.M.

Instructions

- **This is a closed book exam.**
- **The exam comprises one title page, three pages of questions and two pages of formulas.**
- **Answer all six questions in the examination booklets provided.**
- **Calculators and translation dictionaries are permitted.**
- **A formula sheet is provided.**

Good Luck!

Problem 1

Suppose that U is a uniform random variable on the interval $(0, 1)$.

- (a) Find the probability density function of the random variable

$$X = \mu - \beta \ln\{-\ln(U)\},$$

where $\mu \in \mathbb{R}$ and $\beta > 0$ are fixed parameters. **(5 marks)**

- (b) Prove that the moment generating function of X is of the form

$$M_X(t) = e^{\mu t} \Gamma(1 - \beta t).$$

For which values of t does it exist? **(4 marks)**

- (c) Let Y be an arbitrary random variable with moment generating function M_Y . Show that

$$E(Y) = S'_Y(t), \quad \text{Var}(Y) = S''_Y(t),$$

where $S_Y(t) = \ln\{M_Y(t)\}$. **(4 marks)**

- (d) Compute the expectation and variance of X from part (a). Use the fact that $\Gamma'(1) = -\gamma$ and $\Gamma''(1) = \pi^2/6 + \gamma^2$, where $\gamma \approx 0.57722$ is the Euler–Mascheroni constant. **(4 marks)**

Problem 2

Let (X, Y) be a random pair of independent, standard normal random variables. Further, let R and Θ denote the polar coordinates of (X, Y) .

- (a) Show that R has the so-called Rayleigh distribution with density

$$f_R(r) = re^{-r^2/2}, \quad 0 < r < \infty.$$

(5 marks)

- (b) Show that R and Θ are independent. **(5 marks)**

- (c) Derive the distribution of X/Y . What can you say about its moment generating function? **(6 marks)**

Problem 3

The inverse Gaussian distribution with parameters $\chi > 0$ and $\psi > 0$ has probability density function

$$f(x|\chi, \psi) = \frac{\exp \sqrt{\chi\psi}}{\sqrt{2\pi x^3}} \sqrt{\chi} \exp \left\{ -\frac{1}{2}(\chi x^{-1} + \psi x) \right\}, \quad x > 0.$$

- (a) Show that the family $f(x|\chi, \psi)$ is an exponential family. Determine the natural parametrization and the natural parameter space. **(4 marks)**
- (b) Derive expressions for the expectation and variance of random variables $t_1(X), \dots, t_k(X)$ for a k -parameter exponential family with canonical parameters η_1, \dots, η_k and functions $t_1(x), \dots, t_k(x)$ in the usual representation. **(4 marks)**
- (c) Suppose that X is an inverse Gaussian random variable with parameters $\chi > 0$ and $\psi > 0$. Compute $E(1/X)$ and $\text{Var}(1/X)$. **(4 marks)**
- (d) Prove that if X is an inverse Gaussian random variable with parameters $\chi > 0$ and $\psi > 0$,

$$\text{cov}(X, 1/X) = -\frac{1}{\sqrt{\chi\psi}}.$$

(4 marks)

- (e) List three pitfalls of the linear correlation coefficient. **(3 marks)**

Problem 4

- (a) Let X be a random variable such that $X|M = m$ is $\mathcal{N}(m, \sigma^2)$ where $M \sim \mathcal{N}(\mu, \tau^2)$. Determine the distribution of X . **(4 marks)**
- (b) Prove that for any three variables X, Y and Z with finite variances,

$$\text{Cov}(X, Y) = E(\text{Cov}(X, Y|Z)) + \text{Cov}(E(X|Z), E(Y|Z))$$

(4 marks)

- (c) Let X_1 and X_2 be random variables such that $X_i|M = m$ is $\mathcal{N}(m, \sigma^2)$ for $i = 1, 2$ where $M \sim \mathcal{N}(\mu, \tau^2)$. Suppose further that X_1 and X_2 are independent given $M = m$. Compute the correlation between X_1 and X_2 . **(4 marks)**
- (d) Are X_1 and X_2 from part (c) (marginally) independent? Justify your answer. **(4 marks)**

Problem 5

Let X_1, \dots, X_n be a random sample from the Poisson distribution with parameter $\lambda > 0$. Consider

$$Y_n = \sqrt{\frac{1}{n} \sum_{i=1}^n X_i(X_i - 1)}$$

and recall, without proof, the MGF and the mean and variance of the Poisson distribution.

- (a) Determine the exact distribution of $\bar{X}_n = (X_1 + \dots + X_n)/n$. (4 marks)
- (b) Show that $\text{Var}\{X_1(X_1 - 1)\} = 4\lambda^3 + 2\lambda^2$. (4 marks)
- (c) Prove that $E(Y_n) \leq \lambda$. You can use any inequality proved in class. (4 marks)
- (d) Prove that $Y_n \rightarrow \lambda$ in probability. (4 marks)
- (e) Show how the distribution of Y_n can be approximated in terms of the normal distribution when n is large. (4 marks)

Problem 6

Consider i.i.d. random variables X_1, X_2, \dots with density f and distribution function F .

- (a) Prove that the distribution function of $Y_n = \min(X_1, \dots, X_n)$ equals $1 - \{1 - F(x)\}^n$ for all $x \in \mathbb{R}$. (3 marks)
- (b) Suppose that there exists $a \in \mathbb{R}$ such that the density f satisfies $f(x) = 0$ for $x \in (0, a)$ and $f(x) > 0$ for $x \in [a, \infty)$ (in particular, $f(a) > 0$). Prove that $n(Y_n - a)$ converges in distribution to an exponential random variable with parameter $f(a)$. (5 marks)
- (c) Under the conditions of part (b), prove that $Y_n \rightarrow a$ in probability. (4 marks)

DISCRETE DISTRIBUTIONS

	RANGE	PARAMETERS	MASS FUNCTION	CDF	$E_{f_X} [X]$	$\text{Var}_{f_X} [X]$	MGF
	\mathbb{X}		f_X	F_X			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> (r, p)	$\{0, 1, 2, \dots\}$	$r \in \mathbb{Z}^+, p \in (0, 1)$	$\binom{r+x-1}{x} p^r (1-p)^x$		$\frac{r(1-p)}{p}$	$\frac{r(1-p)}{p^2}$	$\left(\frac{p}{1 - e^t(1-p)}\right)^r$

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

$$\Gamma(\alpha) = \int_0^\infty x^{\alpha-1} e^{-x} dx, \quad \alpha > 0$$

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

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

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