

## Topics for the Graduate Exam in Probability (507a)

Most of the following topics are normally covered in the course Math 507a.

This is a one hour exam.

Axioms of probability. Distribution functions. Generating  $\sigma$ -fields. Kolmogorov extension theorem. Borel-Cantelli lemmas.

Random variables. Expectation. Jensen and Chebychev inequalities. Modes of convergence for random variables. Convergence of expected values and moments.

Sums of independent random variables. Weak and strong laws of large numbers. Convergence of random series. Kolmogorov inequality.

Weak convergence, tightness and Helly's Theorem. Characteristic functions. Convolution. Inversion and continuity for characteristic functions. Multidimensional weak convergence and characteristic functions.

The classical Central Limit Theorem. Lindeberg's condition.

Convergence to Poisson.

### References:

P. Billingsley, Probability and Measure  
L. Breiman, Probability  
K.L. Chung, A Course in Probability Theory  
R. Durrett, Probability: Theory and Examples  
A.N. Shiryaev, Probability

## Topics for the Graduate Exam in Statistics 541b

Most of the following topics are normally covered in the course Math 541b.

This is a one hour exam.

Hypotheses testing, Neyman-Pearson lemma, consistency, unbiasedness, power, monotone likelihood ratio, uniformly most powerful tests.

Generalized likelihood ratio procedures, asymptotics

Confidence intervals, tolerance intervals.

Goodness of fit tests, Chi squared test, contingency tables, Kolmogorov Smirnov test.

Sequential testing

The jackknife, jackknife estimate of bias, jackknife estimate of variance, the Efron Stein inequality and applications.

The Bootstrap, smooth bootstrap, bootstrap in regression, bootstrap confidence intervals, bias corrected percentile method confidence intervals.

Theory of Markov chains, stationarity, reversibility.

Hidden Markov models.

Metropolis, Metropolis Hastings algorithm. The Gibbs Sampler.

EM algorithm, asymptotic theory, convergence.

### References:

G. Casella and R.L. Berger, Statistical Inference

T.S. Ferguson, A Course in Large Sample Theory

P.J. Bickel and A. Doksum, Mathematical Statistics

E.L. Lehmann, Theory of Point Estimation

G.J. McLachlan and T. Krishnan, The EM Algorithm and Extensions

B. Efron, The Jackknife, the Bootstrap and Other Resampling Plans

W.R. Gilks, S. Richardson, and D.J. Spiegelhalter, Markov Chain Monte Carlo in Practice

O. Häggström, Finite Markov Chains and Algorithmic Applications

**MATHEMATICAL PROBABILITY AND STATISTICS (II) QUALIFYING  
EXAM (MATH 541A AND 507A)**

FALL 1994

- (1) (a) Define “The family of random variables  $\{X_1, X_2, \dots\}$  is uniformly integrable.”  
 (b) Give an example of a sequence of random variables  $X_n$ ,  $n = 1, 2, \dots$  where  $X_n \geq 0$ ,  $\mathbb{E}X_n = 1$  and the family of random variables  $\{X_1, X_2, \dots\}$  is not uniformly integrable.

For parts (c) and (d) assume that  $X_n \rightarrow X$  a.s. as  $n \rightarrow \infty$  and that  $X_n \geq 0$  for all  $n, \omega$ .

- (c) Assume that  $\{X_1, X_2, \dots\}$  is uniformly integrable. Prove that  $\mathbb{E}X_n \rightarrow \mathbb{E}X$  as  $n \rightarrow \infty$ .  
 (d) Assume that  $\mathbb{E}X_n \rightarrow \mathbb{E}X$  as  $n \rightarrow \infty$ . Prove that  $\{X_1, X_2, \dots\}$  is uniformly integrable.  
 (e) Let  $f : [0, \infty) \rightarrow [0, \infty)$  with

$$\lim_{x \rightarrow \infty} \frac{f(x)}{x} = \infty$$

If  $X_n \geq 0$ ,  $\mathbb{E}f(X_n) \geq c < \infty$ , show that  $\{X_1, X_2, \dots\}$  is uniformly integrable.

- (2) (a) Let  $Y_n$ ,  $n \geq 1$  be random variables and  $Y'_n$  independent of  $Y_n$  with the same distribution as  $Y_n$ . If  $Y_n$  converges in distribution show that  $Y_n \rightarrow Y'_n$  also converges in distribution.  
 (b) Let  $X_1, X_2, \dots$  be independent and identically distributed with characteristic function  $f(t) = \exp(-c|t|^\alpha)$  where  $c > 0$ ,  $\alpha > 0$  are constants, and let  $S_n = X_1 + X_2 + \dots + X_n$ . Find constants  $a_n$  so  $a_n S_n$  converges in distribution to some random variable  $Z$ . How is  $Z$  related to  $X_1$ ?  
 (3) Let  $\theta \in \Theta = (-\infty, \infty)$  and  $X_1, X_2, \dots, X_n$  be independent and identically distributed with density

$$f(x; \theta) = \mathbf{I}(|x - \theta| \leq 1/2)$$

as usual, denote the order statistics by  $X_{(1)} < X_{(2)} < \dots < X_{(n)}$ .

- (a) Show that  $(X_{(1)}, X_{(n)})$  is sufficient for  $\theta$ .  
 (b) Show that  $T_n = \frac{1}{2}(X_{(n)} + X_{(1)})$  is unbiased for  $\theta$ .  
 (c) Compute the variance of  $T_n$ .  
 (d) Find a maximum likelihood estimate for  $\theta$  based on  $(X_{(1)}, X_{(n)})$ . Is it unique?  
 (4) Let  $\theta \in \Theta = (0, \infty)$  and  $X_1, X_2, \dots, X_n$  be independent and identically distributed with density

$$f(x; \theta) = \mathbf{I}(x \in [0, \theta])$$

Construct uniformly minimum variance unbiased estimators  $q(\theta)$  for the following choices of  $q(\theta)$ , or prove they do not exist.

- (a)  $q(\theta) = \theta^k$  for  $k \in \{1, 2, \dots\}$ .  
 (b)  $q(\theta) = e^\theta$ .

MATHEMATICAL PROBABILITY AND STATISTICS (II) QUALIFYING  
EXAM (MATH 541A AND 507A)

SPRING 1995

- (1) Recall that we say the distribution functions  $F_n$  converge in distribution to the distribution function  $F$ , written  $F_n \Rightarrow F$ , if

$$\lim_{n \rightarrow \infty} F_n(x) = F(x)$$

at all continuity points of  $F$ . Show that  $F_n \Rightarrow F$  if and only if for all bounded continuous functions  $h$

$$\int h \, dF_n \rightarrow \int h \, dF \quad \text{as } n \rightarrow \infty$$

- (2) Suppose  $X, X_1, X_2, \dots$  are iid with values in  $(1, \infty)$ . State a condition on  $X$  which is necessary and sufficient for  $\lim_{n \rightarrow \infty} (X_1 X_2 \cdots X_n)^{1/n}$  to exist almost surely and be finite. Demonstrate why your condition is necessary.
- (3) (a) Let  $X_1, \dots, X_n$  be independent normal variates with common variance  $\sigma^2$ . With  $\mu_1, \dots, \mu_n$  not all zero, derive the level  $\alpha$  most powerful test for the hypotheses
- $$H_0 : \mathbb{E}X_1 = \cdots = \mathbb{E}X_n = 0 \quad \text{versus} \quad H_1 : \mathbb{E}X_1 = \mu_1, \dots, \mathbb{E}X_n = \mu_n$$
- (b) Find the level  $\alpha$  most powerful test for the above hypotheses when  $\mathbf{X} = (X_1, \dots, X_n)$  is multivariate normal with known covariance matrix  $\Sigma$ .
- (4) (a) Complete the following statement of the factorization theorem. In a regular model, a statistic  $T(\mathbf{X})$  is sufficient for  $\theta$  if and only if there exists functions  $g$  and  $h$  such that:
- (b) Let  $X_1, X_2, \dots, X_n$  be independent Cauchy ( $\theta$ ) random variables each with density

$$p(x; \theta) = \frac{1}{\pi} \left( \frac{1}{1 + (x - \theta)^2} \right)$$

Show that the order statistics  $(X_{(1)}, \dots, X_{(n)})$  are *minimal* sufficient for  $\theta$ .



MATH 507a/541 QUALIFYING EXAM--SPRING 1997

To pass you must do well enough on both the Probability (problems 1,2,3) and the Statistics (problems 4,5)--high performance on one portion does not compensate for insufficient performance on the other.

(1) Suppose  $X_1, X_2, \dots$  are iid.

(a) If  $E|X_1|^\alpha$  is finite for some  $\alpha > 0$ , show that  $\max_{1 \leq k \leq n} |X_k| / n^{1/\alpha} \rightarrow 0$  a.s.

(b) If  $EX_1$  is finite and nonzero, show that  $\max_{1 \leq k \leq n} |X_k| / |S_n| \rightarrow 0$  a.s.

(2) Suppose  $X_n \rightarrow X$  in distribution and  $Y_n \rightarrow 1$  in probability. Show that  $X_n Y_n \rightarrow X$  in distribution.

(3) Suppose  $\varphi_n$  and  $\varphi_\infty$  are characteristic functions and suppose  $\varphi_n \rightarrow \varphi_\infty$  pointwise, the corresponding d.f.'s  $F_n$  and  $F_\infty$  are continuous, and  $\varphi_n \in L^1$  for all  $n \leq \infty$ .

(a) Use the general inversion formula for characteristic functions (not other methods) to show that the corresponding random variables have densities  $f_n, f_\infty$  given by

$$f_n(x) = (2\pi)^{-1} \int_{-\infty}^{\infty} e^{-itx} \varphi_n(t) dt, \quad n \leq \infty.$$

(b) Let  $\varepsilon > 0$ . Show that if  $h$  is sufficiently small then for all sufficiently large  $n$  (including  $n = \infty$ ),  $\sup_t |\varphi_n(t+h) - \varphi_n(t)| < \varepsilon$ .

It follows from (b) and the Arzela-Ascoli theorem that  $\varphi_n \rightarrow \varphi_\infty$  uniformly on bounded intervals. YOU NEED NOT PROVE THIS, but you may use it.

(c) Suppose  $\varphi_n$  and  $\varphi_\infty$  are all dominated by a function  $g$  in  $L^1$  (that is,  $|\varphi_n(t)| \leq g(t)$  for all  $n$  and all  $t$ .) Show that  $f_n \rightarrow f_\infty$  uniformly on  $\mathbb{R}$ .

4. Let  $\mathbf{X} = (X_1, X_2)$  be bivariate normal, with common unknown mean  $\mu$ , and known variances  $\sigma_1^2$ ,  $\sigma_2^2$ , and correlation  $\rho$ .

a) What is the Fisher information  $\mathbf{I}(\mu)$  based on one observation of the pair  $(X_1, X_2)$ ?

b) Express  $\mathbf{I}(\mu)$  as a function of  $\rho$  in the special case  $\sigma_1^2 = \sigma_2^2 = 1$ .

c) Explain why the expression found in part b takes on the values it does, in the (further) special cases when  $\rho = 0$  and  $\rho = 1$ .

d) What is the Fisher information  $\mathbf{I}(\mu)$  based on one observation  $\mathbf{X}$  of a multivariate  $p$ -dimensional normal vector whose components have common mean  $\mu$  and known covariance matrix  $\Sigma$ ?

5. Recall for  $\mu > 0$  the exponential density with parameter  $\mu$  is  $\mu \exp(-\mu x)$  when  $x$  is positive. Let  $\mu, \nu$  be positive and suppose that  $X_1, \dots, X_n$  are exponential with parameter  $\mu$ , and  $Y_1, \dots, Y_m$  are exponential with parameter  $\nu$  and that all variables are independent.

a) Construct the generalized likelihood ratio test for the hypotheses  $H_0 : \mu = \nu$  versus  $H_1 : \mu \neq \nu$  and show that it can be based on the statistic

$$T(X_1, \dots, X_n, Y_1, \dots, Y_m) = \frac{X_1 + \dots + X_n}{X_1 + \dots + X_n + Y_1 + \dots + Y_m}.$$

In particular, express the critical region of this test in terms of  $T$ .

b) What is the distribution of  $T$  under the null hypotheses?

c) For the case  $n = m$ , show that the rejection region can be written as  $\{T : |T - a| > b\}$  and find  $a$ . Can the type I error probability in this case be written entirely in terms of the cumulative distribution function of  $T$ ?



**M507A and M541B Qualifying Exam.**  
**November 17, 1997**

Answer the Probability section and Statistics section on separate pages.  
JUSTIFY YOUR ANSWERS.

**Section I. Probability**

DO ANY TWO OF THE FOLLOWING FOUR PROBLEMS.

**Q1** (a) For arbitrary events  $A_1, A_2, \dots$ , prove that

$$\mathbb{P}(\liminf A_n) \leq \liminf \mathbb{P}(A_n) \leq \limsup \mathbb{P}(A_n) \leq \mathbb{P}(\limsup A_n).$$

(b) Show, by a single example, that all three inequalities above may be strict.

(c) Prove  $\mathbb{P}(A_n \text{ infinitely often}) = 1$  if and only if  $\sum \mathbb{P}(A_n \cap B) = \infty$  for all  $B$  with  $\mathbb{P}(B) > 0$ .

**Q2** Assume  $X_1, X_2, \dots$  are random variables, not necessarily independent, and not necessarily identically distributed. Assume  $\mathbb{E}X_i = 0$  for all  $i$  and  $\mathbb{E}(X_i X_j) \leq 0$  for all  $i \neq j$ . Assume that there is a finite constant  $K$  such that  $\mathbb{E}X_i^2 \leq K$  for all  $i$ . Write  $S_n := X_1 + X_2 + \dots + X_n$ . For some fixed  $\epsilon > 0$ , write  $A_n$  for the event  $\{|S_n| \geq n\epsilon\}$ .

(a) Prove that  $\text{Var}(S_n) = O(n)$ .

(b) Using a), prove that for any  $\epsilon > 0$ ,  $\mathbb{P}(A_n) = O(1/n)$ .

(c) Let  $B_n = A_{n^2}$ . Prove that  $\mathbb{P}(B_n \text{ i.o.}) = 0$ .

(d) Analyze the random variable  $D_n := \max_{n^2 \leq k < (n+1)^2} |S_k - S_{n^2}|$  to show  $n^{-2}D_n \rightarrow 0$  with probability 1.

**Q3** Let  $X, X_1, X_2, \dots$  be i.i.d. with characteristic function  $\phi(t) := \mathbb{E}e^{itX}$ . Write  $S_n = X_1 + \dots + X_n$ .

(a) Prove that if  $\phi'(0) = ia$  then  $S_n/n \rightarrow a$  in probability.

(b) Justify the claim: if  $S_n/n \rightarrow a$  in probability, then  $\phi'(0)$  exists and equals  $ia$ . You may assume without proof that for characteristic functions, if  $\phi_n \rightarrow \phi$  pointwise, then the convergence is uniform on compact sets.

**Q4** Assume  $X, X_1, X_2, \dots$  are i.i.d. with  $X \geq 0$  always.

(a) Prove that if  $\mathbb{E}X < \infty$  then  $X_n/n \rightarrow 0$  almost surely.

(b) Prove that if  $\mathbb{E}X = \infty$ , then  $\limsup X_n/n = \infty$  almost surely, and hence  $(X_1 + \dots + X_n)/n \rightarrow \infty$  almost surely.

## Section II. Statistics

DO ANY TWO OF THE FOLLOWING THREE PROBLEMS.

**Q1** Let  $X_1, X_2, \dots, X_n$  be independent and identically distributed random variables having exponential distribution with mean  $1/\theta$ .

(a) Let  $X_{(1)} \leq \dots \leq X_{(n)}$  denote the order statistics, and write  $T_1 = X_{(1)}, T_2 = X_{(2)} - X_{(1)}, \dots, T_k = X_{(k)} - X_{(k-1)}$  for  $2 \leq k \leq n$ . Find the joint distribution of  $T_1, T_2, \dots, T_k$ .

(b) Light bulbs are known to have this exponential distribution. Suppose that a random sample of  $n$  bulbs is put under observation, and observation is stopped when the  $k^{\text{th}}$  bulb burns out. Find the maximum likelihood estimator of the mean lifetime of a bulb.

(c) Find a 95% confidence interval for this mean.

**Q2** (a) Give the definition of a sufficient statistic.

(b) Let  $X_1, \dots, X_n$  be a random sample from a Poisson distribution with parameter  $\lambda$ . Find a sufficient statistic for  $\lambda$ .

(c) Show that  $T = \frac{1}{n} \sum_{j=1}^n I(X_j = 0)$  is an unbiased estimator of  $e^{-\lambda}$ , and find its variance. Here,  $I(A) = 1$  if  $A$  is true,  $= 0$  if false.

(d) Use the Rao-Blackwell Theorem and (c) to find a better estimator of  $e^{-\lambda}$ .

(e) What optimality properties does the estimator in (d) have?



- Q3 (a) State the Cramér-Rao Inequality.  
 (b) Let  $X_1, \dots, X_n$  be independent and identically distributed Bernoulli random variables with mean  $\theta$ . Find the Cramér-Rao lower bound for the variance of unbiased estimators of  $\tau(\theta) = \theta(1 - \theta)$ .  
 (c) State and prove the Neyman-Pearson Lemma.  
 (d) 1000 individuals were classified according to sex, and according to whether or not they were color-blind as follows:

	Male	Female
Normal	442	514
Color-blind	38	6

According to a genetic model, these numbers should have relative frequencies given by

$$\frac{p/2}{q/2} \mid \frac{p^2/2 + pq}{q^2/2}$$

where  $q = 1 - p$  is the proportion of color-blind individuals in the population. Are the data consistent with this model?

MATH 507a/541b QUALIFYING EXAM – FALL 1998

To pass you must do well enough on both the Probability and the Statistics parts – high performance on one portion does not compensate for low performance on the other.

STATISTICS

1. There are  $g$  categories,  $i = 1, 2, \dots, g$  with probabilities  $\pi_1, \pi_2, \dots, \pi_g$ . The random variable  $X$  is defined by

$$P(X = i) = \pi_i, \quad 1 \leq i \leq g.$$

Let  $X_1, X_2, \dots, X_n$  be iid as  $X$ . Define  $Z_{ij}$  by

$$Z_{i,j} = \mathbf{I}(X_j = i).$$

The number of times the variables  $X_1, \dots, X_n$  fall in category  $i$  is given by

$$Y_i = \sum_{j=1}^n Z_{ij}.$$

- a) Find the mean and variance of  $Y_1$ .
- b) Find  $P((Y_1, \dots, Y_g) = (n_1, \dots, n_g))$ .
- c) Find the mean vector  $\mu$  and covariance matrix  $\Sigma$  of the vector  $(Y_1, \dots, Y_g)$ .
- d) Give a large sample test for the hypothesis  $H_0 : \pi_1 = \pi_2$ .

2. Let  $X_1, \dots, X_n$  be iid  $\mathcal{N}(\mu, \sigma^2)$  with known mean  $\mu$ .

a) Find the UMVE of the parameter  $\sigma^2$  and prove it to be such.

b) Of *all* estimators of  $\sigma^2$  of the form

$$\hat{\theta} = a \sum_{i=1}^n (X_i - \mu)^2, \quad a \in \mathbf{R},$$

find the one which achieves the smallest mean square error. (Hint: If  $Z$  is  $\mathcal{N}(0, 1)$  then  $EZ^4 = 3$ .)



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507a Qualifying exam. November 1998

Do two of the following three problems. If you hand in more than two, only the first two, in order of the given problem numbers, will be graded!

1. Let  $E_1, E_2, \dots$  be arbitrary events. Let  $G := \limsup_n E_n$ . Show that  $\mathbb{P}(G) = 1$  if and only if  $\sum_n \mathbb{P}(A \cap E_n) = \infty$  for all events  $A$  having  $\mathbb{P}(A) > 0$ .

2.) Assume that  $A_i$  are independent events, and let  $X_i$  be the indicator random variable for the event  $A_i$ . Let  $f(n) := n^{-1} \sum_1^n \mathbb{P}(A_i)$ , and write  $S_n := \sum_1^n X_i$ . Prove that  $S_n/n - f(n)$  converges to zero in probability.

3). a) For  $X_\lambda$  having the Poisson distribution with mean  $\lambda > 0$ , show that  $(X_\lambda - \lambda)/\sqrt{\lambda}$  converges in distribution to the standard normal, as  $\lambda \rightarrow \infty$ . Do not restrict to integer valued  $\lambda$ .

b) For  $X, X_1, X_2, \dots$  i.i.d. with the symmetric density

$$f(x) = c \frac{1}{x^2 \log |x|} \text{ for } |x| > 4$$

with the appropriate normalizing constant  $c$ ,

b1) show that the characteristic function  $\phi$  for  $X$  has  $\phi'(0) = 0$ , and

b2) show that this implies that  $(X_1 + \dots + X_n)/n$  converges to zero in probability.

b3) Also show that  $\mathbb{E}|X| = \infty$ .

507a Qualifying exam. May 17, 1999

1.) Assume that  $X_1, X_2, \dots$  are independent, and take values in a countable set  $A \subset (0, \infty)$ . Assume that there are constants  $c_1, c_2, \dots > 0$  such that

$$\sum_1^\infty \mathbb{P}(X_i \neq c_i) < \infty, \text{ and } \sum_1^\infty c_i < \infty.$$

Let  $S = \sum_1^\infty X_i$ . Prove that  $S$  is a discrete random variable, i.e., there is a countable set  $B \subset (-\infty, \infty)$  such that  $1 = \mathbb{P}(S \in B)$ .

2.) Assume that  $X_1, X_2, \dots$  are independent, with  $\mathbb{P}(X_k = 1) = 1/k$  and  $\mathbb{P}(X_k = 0) = 1 - 1/k$ . Let  $S_n = X_1 + \dots + X_n$ , so that  $h(n) := \mathbb{E}S_n = 1 + (1/2) + \dots + (1/n)$ , with  $h(n) \sim \log n$  as  $n \rightarrow \infty$ . The goal is to show, USING characteristic functions, that  $(S_n - h(n))/\sqrt{h(n)}$  converges in distribution to the standard normal random variable  $Z$  with mean 0 and variance 1.

a) What is the characteristic function  $\phi$  of the standard normal, namely  $\phi(u) = \mathbb{E}e^{iuZ}$ , in simplified form?

b) Give explicitly the characteristic function  $\phi_k$  of the mean zero random variable  $X_k - 1/k$ , namely  $\phi_k(u) = \mathbb{E}e^{iu(X_k - 1/k)}$ .

c) Show that for each  $k$ , as  $t \rightarrow 0$

$$\phi_k(t) = 1 - \frac{k-1}{k^2} \frac{t^2}{2} + o(t^2).$$

d) Write  $\phi_n^*$  for the characteristic function of  $(S_n - h(n))/\sqrt{h(n)}$ . Express  $\phi_n^*(u)$  in terms of the functions  $\phi_k$ .

e) Show that, for fixed  $u$ , as  $n \rightarrow \infty$ ,  $\phi_n^*(u) \rightarrow \phi(u)$ .



1. Let  $a_n$  and  $\mu_n$  be deterministic sequences tending to  $\infty$  and  $\mu$  respectively, and assume that the random variables  $X_n$ , properly scaled, converge in distribution to  $X$ ; in particular, that

$$a_n(X_n - \mu_n) \xrightarrow{d} X.$$

a) Prove that if  $g$  is a function having a continuous derivative at  $\mu$ , then

$$a_n(g(X_n) - g(\mu_n)) \xrightarrow{d} g'(\mu)X.$$

Now let  $Y_1, \dots, Y_n$  be a sample of independent exponential variables ('failure times') with density  $f(t; \lambda) = \lambda e^{-\lambda t}$  for  $\lambda, t$  positive.

b) Calculate the Fisher information for  $\lambda$  in the sample.

c) Find, and justify, the limiting distribution of the maximum likelihood estimator for  $\lambda$ .

d) Suppose it is desired to estimate the probability that an exponential from the same distribution will not fail before time  $x$ ; that is, we wish to estimate

$$q(\lambda) = P(Y > x) = e^{-\lambda x}.$$

What is the limiting distribution of the maximum likelihood estimator of  $q(\lambda)$ ? (Hint: Use part a)

2. a) Prove the following form of the Neyman Pearson Lemma: If  $\mathbf{X} \in \mathbf{R}^n$  is a random vector with density  $f(\mathbf{x}; \theta)$ , where  $\theta \in \{\theta_0, \theta_1\}$ , then the test for  $H_0 : \theta = \theta_0$  versus  $H_1 : \theta = \theta_1$  which rejects  $H_0$  when  $L(\mathbf{X}) = f(\mathbf{x}; \theta_1)/f(\mathbf{x}; \theta_0)$  exceeds a level  $k$  achieves the maximum power over all tests of size  $P_0(L(\mathbf{X}) \geq k)$ .

b) Let  $X_1, \dots, X_n$  be independent exponential variables with parameters either  $\mu_1, \dots, \mu_n$ , or  $\nu_1, \dots, \nu_n$ , known values. Find a simple test and test statistic for the Neyman Pearson tests that distinguish between the two hypotheses.



507a Qualifying exam. December 17, 1999

1.) Let  $(X_1, X_2, \dots)$  and  $(X'_1, X'_2, \dots)$  have the same distribution in  $\mathbb{R}^\infty$ . Prove that if  $X_n \rightarrow X$  a.s., then there exists a random variable  $X'$  such that  $X'_n \rightarrow X'$  a.s.

2.) Show that if  $X, Y$  are independent random variables and the distribution of  $X$  is absolutely continuous, then so is the distribution of  $X + Y$ .

3.) Let  $X_k$  have characteristic function  $\phi_k$ , with  $X_1, X_2, \dots$  independent. Show that  $\sum_1^n X_k$  converges almost surely *if and only if* there exists a neighborhood  $U$  of 0 and a function  $h$  with  $\prod_1^n \phi_k(u) \rightarrow h(u) \neq 0$  for all  $u \in U$ . [Hint: the *only if* is easy. For the *if*, consider the characteristic functions of the partial sums  $\sum_m^n X_k$ .]

4.) Let  $Z, Z_1, Z_2, \dots$  be iid with  $\mathbb{P}(Z = -1) = \mathbb{P}(Z = 1) = 1/2$ , and let constants  $c_1, c_2, \dots$  be given.

a) Express the characteristic function of  $\sum_1^n c_k Z_k$  in terms of standard elementary functions.

b) Assume the result you were asked to prove in 3). Show that  $\sum_{k \geq 1} c_k Z_k$  converges almost surely *if and only if*  $\sum c_k^2 < \infty$ . [If you use some tool other than problem 3, be sure to fully state the result you are using.]

### Math 541 Exam Portion

1.a) Let  $a_n$  and  $\mu_n$  be deterministic sequences tending to  $\infty$  and  $\mu$  respectively, and assume that the random variables  $X_n$ , properly scaled, converge in distribution to  $X$ ; in particular, that

$$a_n(X_n - \mu_n) \xrightarrow{d} X.$$

Prove that if  $g$  is a function having a continuous derivative at  $\mu$ , then

$$a_n(g(X_n) - g(\mu_n)) \xrightarrow{d} g'(\mu)X.$$

b) State a multidimensional version of this fact.

Now let  $X_1, \dots, X_n$  be iid with mean  $\mu$  and variance  $\sigma^2$ .

c) Find a method of moments estimator for the coefficient of variation

$$CV = \frac{\sigma}{\mu}.$$

d) Find the asymptotic distribution of the estimator in c). What moments of the  $X$  distribution need to exist?

2) Let  $X_1, \dots, X_n$  be iid normal with unknown mean  $\mu$  and known variance  $\sigma^2$ .

a) Find the critical region for the Neyman Pearson test at level  $\alpha \in (0, 1)$  for  $H_0: \mu = \mu_0$  versus  $H_1: \mu = \mu_1$  with  $\mu_0 < \mu_1$ .

b) Determine the power function  $\beta(\mu)$  of this test.



## PROBABILITY SECTION

**Q1.** Let  $A_1, A_2, \dots$  be (possibly dependent) events, let

$$e_n := \sum_{1 \leq i \leq n} \mathbb{P}(A_i), \quad f_n := \sum_{1 \leq i, j \leq n} \mathbb{P}(A_i \cap A_j),$$

with  $e_n \rightarrow \infty$  as  $n \rightarrow \infty$ .

(a) Give the definition of the event  $G = \{A_n \text{ i.o.}\}$  in terms of union and intersection.

(b) Give an example of the above, in which  $\mathbb{P}(A_i \text{ i.o.}) = 0$ , and in which you can explicitly calculate the  $e_n, f_n$  to show  $f_n/e_n^2 \rightarrow \infty$ .

(c) Now assume that as  $n \rightarrow \infty$  we have

$$\beta := \lim f_n/e_n^2 \text{ exists, with } \beta < \infty.$$

Show that  $\mathbb{P}(A_n \text{ i.o.}) \geq 1/\beta > 0$ .

[Hint: consider  $X_n := \sum_{1 \leq i \leq n} 1_{A_i}$ , so that  $e_n = \mathbb{E}X_n$  and  $f_n = \mathbb{E}X_n^2$ . Consider  $Y_n := X_n/e_n$ . For  $\epsilon > 0$  let  $Z_n$  be the indicator of the event that  $Y_n \geq \epsilon$ . Show that  $\mathbb{E}(Y_n Z_n) \geq 1 - \epsilon$ . Apply Cauchy-Schwarz to  $Y_n Z_n$ .]

**Q2.** Assume that  $X, X_1, X_2, \dots$  are i.i.d., and write  $S_n = X_1 + \dots + X_n$ . Assume that  $\mathbb{E}X = 0$  and  $\mathbb{E}X^2 = \sigma^2 \in (0, \infty)$ .

(a) What can be said, relevant to part (b), about the expansion of  $\phi(t) := \mathbb{E}e^{itX}$ , as  $t \rightarrow 0$ ?

(b) Give the statement of the Central Limit Theorem for  $S_n$ , AND give a sketch or outline of the proof using characteristic functions.



MATH 507a/541 QUALIFYING EXAM. MAY 15, 2000.

PLEASE NOTE: To pass you must do well enough on both the Probability and the Statistics sections. High performance in one portion does not compensate for insufficient performance on the other.

STATISTICS SECTION

**Q1.** Let  $X_1, X_2, \dots, X_n$  be independent and identically distributed random variables having density

$$f(x) = \begin{cases} e^{-(x-\theta)} & \text{if } x > \theta \\ 0 & \text{if } x \leq \theta \end{cases}$$

for some  $\theta \in (-\infty, \infty)$ .

- (a) State the Factorization Theorem for sufficient statistics.
- (b) Find a one-dimensional sufficient statistic for  $\theta$ .
- (c) Find a 95% confidence interval for  $\theta$ .
- (d) Derive the likelihood ratio test of the null hypothesis that  $\theta \geq 0$  against the alternative that  $\theta < 0$ .

**Q2.** Outputs  $X_1, X_2, \dots, X_n$  from a physical device are independent and identically distributed random variables having exponential distribution with (unknown) mean  $\lambda^{-1}$ . A measuring device records the values of the  $X_j$  as long as  $X_j < c$ , for some known threshold  $c > 0$ . If  $X_j \geq c$  then the device returns the value  $c$ . Define

$$S_n = \sum_{j=1}^n X_j I(X_j < c), \quad T_n = \sum_{j=1}^n I(X_j \geq c),$$

where  $I(A)$  denotes the indicator of the event  $A$ .

- (a) Write down the likelihood function of the observed values in terms of  $S_n$  and  $T_n$ .

(b) Show that the Maximum Likelihood Estimator of  $\lambda$  is

$$\hat{\lambda} = \frac{n - T_n}{S_n + cT_n}.$$

(c) Find the joint asymptotic distribution of  $(S_n, T_n)$ .

Hint:

$$\int_0^c x \lambda e^{-\lambda x} dx = \lambda^{-1} (1 - (1 + c\lambda)e^{-c\lambda})$$

and

$$\int_0^c x^2 \lambda e^{-\lambda x} dx = \lambda^{-1} (2 - (2 + 2c\lambda + c^2\lambda^2)e^{-c\lambda}).$$

(d) Using the result of the previous part, or otherwise, find the asymptotic distribution of  $\hat{\lambda}$ .

507a Qualifying exam. May 8, 2001

1.) Suppose  $X, X_1, X_2, \dots$  are iid with  $\mathbb{E}|X| = \infty$ . Let  $S_n = X_1 + \dots + X_n$  and let  $M_n = S_n/n$ . Let  $A$  be the event that  $M_n$  converges to a finite limit. Let  $B$  be the event that  $|X_n| \geq n$  infinitely often.

- a.) State the definition of  $B$  in terms of unions and intersections.
- b.) Show that  $\mathbb{P}(B) = 1$ .
- c.) Use

$$M_n - M_{n+1} = M_n/(n+1) - X_{n+1}/(n+1)$$

to show that  $A \cap B = \emptyset$ .

- d.) Complete the proof that  $\mathbb{P}(A) = 0$ .

2.) Let  $M(t) = \mathbb{E}e^{tX}$  be the moment generating function of a random variable  $X$ . Let  $I = \{t \in (-\infty, \infty) : M(t) < \infty\}$ .

- a.) Show that  $I$  is an interval.
- b.) Show that  $M$  is continuous on the interior of  $I$ .
- c.) Give an example where  $I = (-\infty, 1)$ .

3.) Let  $X_1, X_2, \dots$  be independent, with

$$\mathbb{P}(X_n = 1) = \mathbb{P}(X_n = -1) = \frac{1}{2} \left(1 - \frac{1}{n^2}\right)$$

and

$$\mathbb{P}(X_n = n) = \mathbb{P}(X_n = -n) = \frac{1}{2n^2}.$$

Let  $S_n = X_1 + \dots + X_n$  and  $S_n^* = S_n/\sqrt{n}$ .

- a) Show that  $\text{VAR}(S_n^*) \rightarrow 2$ .

b) Show that  $S_n^* \Rightarrow Z$  where  $Z$  is normal; you pick the parameters for  $Z$ . Even if you cannot this distributional convergence, state the mean and variance of the limit random variable  $Z$ . *prove*



Math 541a Exam Portion. Spring 2001

**Problem 1.** a) Let  $\mathbf{X} \sim \mathcal{N}(\lambda\mu, v^2\Sigma)$ , where  $\mu \in \mathbf{R}^n$  and  $\Sigma \in \mathbf{R}^{n \times n}$  are a known vector and positive definite matrix respectively, and  $\lambda \in \mathbf{R}$  and  $v^2 > 0$  are unknown parameters in  $\mathbf{R}$ .

- a) Find the maximum likelihood estimators  $\hat{\lambda}$  and  $\hat{v}^2$  of  $\lambda$  and  $v^2$  on the basis of the observation  $\mathbf{X}$ .
- b) Determine whether or not  $\hat{\lambda}$  is unbiased for  $\lambda$ .
- c) Calculate the variance of  $\hat{\lambda}$ .
- d) *Demonstrate* what the estimators  $\hat{\lambda}$  and  $\hat{v}^2$  become when  $\mu = (1, 1, \dots, 1)$  and  $\Sigma$  is the identity matrix. Explain.

**Problem 2.** a) Let  $\theta > 0$  be unknown and suppose that  $(X, Y)$  is uniform over the triangular region with vertices at  $(0, 0)$ ,  $(\theta, 0)$  and  $(0, \theta)$ . Let  $(X_i, Y_i)$  be iid as  $(X, Y)$ .

- a) Find a one dimensional sufficient statistic  $T$  for  $\theta$ , and prove it is sufficient.
- b) Find an unbiased estimate of  $\theta$  which is a function of  $T$ .
- c) Is  $\hat{\theta}$  UMVU? Prove your claim.

507a Qualifying exam. September 12, 2001. Be sure to attempt the later parts of each problem even if you cannot do one of the earlier parts.

1.)

a) Prove that for any sequence  $X_n$  of random variables there exist positive constants  $c_n$  such that  $X_n/c_n$  converges to 0 almost surely.

b) Can you choose  $c_n$  so that this convergence is pointwise at every  $\omega \in \Omega$ ?

c) Now suppose that  $X_1, X_2, \dots$  are iid, that  $\mathbb{E}X_1$  exists and is finite, and that  $c_n = n$ . Prove that  $X_n/c_n$  converges to 0 almost surely.

2.) Assume  $X, X_1, X_2, \dots$  i.i.d. with characteristic function  $\phi$  for  $X$ , i.e.  $\phi(t) := \mathbb{E}e^{itX}$ , and let  $S_n := X_1 + \dots + X_n$

a) For a random variable  $X$ , what special property of its characteristic function  $\phi$  holds if and only if  $X$  and  $-X$  have the same distribution? (Show both the implications.)

b) Express the characteristic function of the sample average,  $\phi_{S_n/n}(t)$ , in terms of  $\phi$ .

c) If  $X$  has  $\phi'(0) = 0$ , show that  $(X_1 + \dots + X_n)/n$  converges to zero in probability. [HINTS: Since  $\phi(0) = 1$ ,  $\phi'(0) = 0$  if and only if  $\phi(u) = 1 + o(u)$  as  $u \rightarrow 0$ . Also, for fixed  $t$ , as  $n \rightarrow \infty$ ,  $(1 + o(t/n))^n \rightarrow 1$  can be shown from  $\log(1+x) \sim x$  for small positive  $x$ .]

From now on assume that  $X, X_1, X_2, \dots$  are i.i.d. with the symmetric density

$$f(x) = c \frac{1}{x^2 \log|x|} \text{ for } |x| > 4; \quad f(x) = 0 \text{ otherwise,}$$

where  $c$  is an appropriate normalizing constant.

d) Show that  $\mathbb{E}|X| = \infty$ .

e) Show that the characteristic function  $\phi$  for  $X$  has  $\phi'(0) = 0$ . [HINT: express  $1 - \phi(t)$  as an integral over  $x > 4$  and use the change of variables  $y = tx$  to show that  $|1 - \phi(t)|/t \rightarrow 0$  as  $t \rightarrow 0$ . You might use  $|1 - \cos y| \leq y^2 \forall y$ , and dominated convergence.]

February 2002

**Math 507a Exam**

**Problem 1.**

Let  $X_1, X_2, \dots$  be a sequence of iid random variables so that  $EX_1 = 0, E|X_1|^2 < \infty$ . Show that

$$\lim_{n \rightarrow \infty} \max \left\{ \frac{|X_1|}{\sqrt{n}}, \dots, \frac{|X_n|}{\sqrt{n}} \right\} = 0$$

in probability.

**Problem 2.**

Let  $X_1, X_2, \dots$  be a sequence of random variables and  $\varepsilon_1, \varepsilon_2, \dots$ , a sequence of real numbers so that  $\varepsilon_n > 0$ ,  $\sum_{n \geq 1} \varepsilon_n < \infty$ , and  $\sum_{n \geq 1} P(|X_n| > \varepsilon_n) < \infty$ . Show that the series  $\sum_{n \geq 1} |X_n|$  converges with probability one.



**Math 507a Qualifying Exam**  
**Fall 2002**

You should try at least 3 problems; you may try all 4.

**Problem 1.** Let  $X_1, X_2, \dots$  be independent r.v.'s such that  $X_n$  is uniformly distributed on  $[-n, n]$  for  $n = 1, 2, \dots$ . Let  $S_n = X_1 + X_2 + \dots + X_n$ . Prove that for some  $\alpha$ ,  $0 < \alpha < \infty$ ,

$$S_n/n^\alpha \longrightarrow Z \quad \text{in distribution,}$$

where  $Z$  is a normal random variable. Identify  $\alpha$  and the parameters of the normal  $Z$ .

**Problem 2.** Let  $Y_1, Y_2, \dots$  be iid non-negative random variables. Let  $S = \sum_{n=1}^{\infty} \alpha^n Y_n$ , where  $0 < \alpha < 1$ .

- a) Show that  $EY < \infty$  implies  $S < \infty$  a.s.
- b) Give an example where  $S = \infty$  a.s.

**Problem 3.** Suppose  $X$  and  $Y$  are independent random variables and for some  $p > 0$  we have  $E|X + Y|^p < \infty$ . Show that  $E|X|^p < \infty$ . HINT: For  $a, b \in \mathbb{R}$ ,  $|a + b|^p \leq 2^p(|a|^p + |b|^p)$ .

**Problem 4.** A *median* of a r.v.  $X$  is a number  $m$  such that

$$P(X \leq m) \geq \frac{1}{2}, \quad P(X \geq m) \geq \frac{1}{2};$$

note  $m$  need not be unique. Show that if  $X_n \rightarrow X_\infty$  in distribution,  $m_n$  is a median of  $X_n$ ,  $m_\infty$  is a median of  $X_\infty$ , the distribution function  $F_\infty$  is continuous, and  $m_\infty$  is unique, then  $m_n \rightarrow m_\infty$ .

Math 541b Qualify Exam. Fall 2002

**Problem 1.** (EM) There are two possibly biased coins. The probability of heads for the first coin is  $1/3$  and the probability of heads in the second coin is  $p \in (0, 1)$ , an unknown parameter. An experiment consists of tossing the two coins together, which we do  $n$  times. Only  $X_i$ , the number of heads in the  $i^{\text{th}}$  experiment, is observable.

1. Let  $n_j$ ,  $j = 0, 1, 2$  be the number of experiments where  $j$  heads show up. Write the joint distribution of  $(X_1, X_2, \dots, X_n)$  in terms of  $n_0, n_1, n_2$ .
2. Write an equation for the maximum likelihood estimate (MLE) of  $p$ . Is it easy to solve this equation? If not, design an expectation-maximization (EM) algorithm for calculating this MLE.
3. Although we do not have a 'closed form' maximum likelihood estimator  $\hat{p}$  for  $p$ , we can still study its approximate distribution. What is the approximate distribution of  $\hat{p}$  when the sample size  $n$  tends to infinity?
4. It was suspected that the second coin is unbiased, that is, that  $p = 1/2$ . Outline a procedure for testing this hypothesis.

**Problem 2.** Let  $\Theta = (0, \infty)$  and suppose that the density  $f(x, y; \theta)$  of  $(X, Y)$  is uniform over region  $A$ , where  $H_0 : A = [-\theta, \theta]^2$  (the square of side length  $2\theta$  centered at the origin, or  $H_1 : A$  is the circle of radius  $\theta$  centered at the origin. Let  $(X_1, Y_1), \dots, (X_n, Y_n)$ , be i.i.d. with density  $f(x, y; \theta)$ .

a) For fixed known  $\theta \in \Theta$ , describe the (non-trivial) Neyman Pearson tests for the testing between the simple hypotheses  $H_0$  vs.  $H_1$ ?

b) A hypotheses test is said to be *consistent* if the probability of rejecting the null hypotheses when it is false tends to 1 as the sample size  $n$  tends to infinity. Prove that the test in part a) is consistent.

c) Describe a consistent test for the composite hypotheses  $H_0$  vs.  $H_1$  when  $\theta$  is only known to lie in  $\Theta$ .



**Math 507a Qualifying Exam**  
**Spring 2003**

You should try at least 3 problems; you may try all 4.

**Problem 1.** Let  $X_1, X_2, \dots$  be iid with characteristic function  $\varphi$ , and suppose  $\varphi'(0) = ia$  for some real  $a$ . Show that  $(X_1 + \dots + X_n)/n \rightarrow a$  in probability.

**Problem 2.** Let  $X_n$  and  $X'_n$  be independent with the same d.f.  $F_n$ , and suppose  $X_n - X'_n \rightarrow 0$  in distribution. Show that there exist constants  $a_n$  such that  $X_n - a_n \rightarrow 0$  in probability. HINT: You can use  $a_n = \inf\{x : F_n(x) \geq 1/2\}$ .

**Problem 3.** Let  $\alpha > 0$  and let  $X_1, X_2, \dots$  be iid with d.f.  $F(x) = 1 - x^{-\alpha}$ ,  $x \geq 1$ . Let  $M_n = \max(X_1, \dots, X_n)$ . Find  $0 < \beta < \infty$  and a nondecreasing sequence of constants  $a_n$  such that

$$\limsup_n \frac{\log M_n}{a_n} = \beta \quad \text{a.s.}$$

HINT: First prove the same thing with  $X_n$  in place of  $M_n$ .

**Problem 4.** Let  $p \geq 1$ , and let  $X_1, X_2, \dots$  be random variables with  $E|X_n|^p < \infty$  for all  $n$ .

(i) If there exists a random variable  $X$  such that  $E|X_n - X|^p \rightarrow 0$  as  $n \rightarrow \infty$ , show that  $E|X_n - X_m|^p \rightarrow 0$  as  $n, m \rightarrow \infty$ .

(ii) If  $E|X_n - X_m|^p \rightarrow 0$  as  $n, m \rightarrow \infty$ , show that  $\{X_n\}$  has a subsequence which converges a.s.

(iii) If  $E|X_n - X_m|^p \rightarrow 0$  as  $n, m \rightarrow \infty$ , show that there exists a random variable  $X$  such that  $E|X|^p < \infty$  and  $E|X_n - X|^p \rightarrow 0$  as  $n \rightarrow \infty$ .



Spring 2003 Math 541b Exam

1. Let  $\theta = (\theta_1, \dots, \theta_k)$  be a vector of given probabilities, and  $\mathbf{e}_i$  the  $i^{\text{th}}$  unit vector in  $\mathbf{R}^k$  with a 1 in position  $i$  and zeros elsewhere. Let  $\mathbf{Y}, \mathbf{Y}_1, \dots, \mathbf{Y}_n$  be i.i.d. with distribution

$$P(\mathbf{Y} = \mathbf{e}_i) = \theta_i, \quad \mathbf{X}_n = \sum_{j=1}^n \mathbf{Y}_j \quad \text{and} \quad \bar{\mathbf{X}}_n = \frac{1}{n} \sum_{j=1}^n \mathbf{Y}_j$$

so that  $\mathbf{X}_n = (n_1, \dots, n_k)$  has the multinomial distribution  $M(n, \theta)$ .

a) Find the mean vector  $\mu$  and covariance matrix  $\Sigma$  of  $\mathbf{Y}$ .

b) Write the usual chi-squared statistic

$$V_n = \sum_{i=1}^k \frac{(n_i - n\theta_i)^2}{n\theta_i}$$

as

$$V_n = n(\bar{\mathbf{X}}_n - \theta)' P^{-1} (\bar{\mathbf{X}}_n - \theta)$$

for some diagonal matrix  $P$ .

c) Find the asymptotic distribution of  $\sqrt{n}(\bar{\mathbf{X}}_n - \theta)$ .

d) Show that as  $n \rightarrow \infty$ ,  $V_n \rightarrow_d \chi_{k-1}^2$ , a chi squared random variable on  $k-1$  degrees of freedom. Hint: Write the asymptotic distribution in terms of a vector with covariance matrix  $\Gamma$  which satisfies  $\Gamma' = \Gamma$  and  $\Gamma^2 = \Gamma$ .

2. Suppose data  $X_1, X_2, \dots, X_n$  are independent identically distributed normal random variables with mean  $\mu$  and variance  $\sigma_0^2$ . Suppose that  $\mu$  is random with (prior) normal distribution  $\mathcal{N}(\mu_0, \sigma_0^2)$ . What is the conditional distribution (posterior) of  $\mu$  given the data? Give the mean and variance of the posterior distribution of  $\mu$  in terms of  $\bar{X}, \mu_0, \sigma_0^2$ .

**Math 507a Qualifying Exam Problems**  
**Fall 2003**

**Problem 1.** Let  $X_1, X_2, \dots$  be nonnegative iid random variables with finite mean. Show that

$$\lim_n \frac{1}{n} E\left(\max_{j \leq n} X_j\right) = 0.$$

**Problem 2.** Let  $X$  be a random variable and  $f : \mathbb{R} \rightarrow \mathbb{R}$  a measurable function. If  $X$  and  $f(X)$  are independent, show  $f$  is constant.

**Problem 3.** Let  $X_1, X_2, \dots$  be iid with mean  $\mu$  and variance  $\sigma^2$ . Let  $N_\lambda$  be  $\text{Poisson}(\lambda)$  independent of the  $X_i$ 's.

(a) Find the limit in distribution as  $\lambda \rightarrow \infty$  for

$$\frac{\sum_{i=1}^{N_\lambda} X_i - N_\lambda \mu}{\sqrt{\lambda}}.$$

(b) Find the limit in distribution as  $\lambda \rightarrow \infty$  for

$$\frac{\sum_{i=1}^{N_\lambda} X_i - \lambda \mu}{\sqrt{\lambda}}.$$

(c) For which random variables  $X$  will the two limits be the same?

**Problem 4.** Let  $X$  and  $Y$  be independent  $N(0, 1)$  random variables, and let  $Z = X + Y$ .

(a) Show that  $E(Z|X > 0, Y > 0) = 2\sqrt{2/\pi}$

(b) Find the distribution and the density of  $Z$  given that  $X > 0, Y > 0$ .

Math 541b, Fall 2003

1. Consider a test with critical region of the form  $\{T \geq c\}$  for testing  $H : \theta = 0$  versus  $K : \theta > 0$ . Suppose that  $T$  has a continuous distribution  $F_\theta$ . Define the  $p$ -value as

$$U = 1 - F_0(T).$$

- a) Show that if the test has level  $\alpha$ , the power is

$$\beta(\theta) = P\{U \leq \alpha\} = 1 - F_\theta(F_0^{-1}(1 - \alpha)),$$

where  $F_0^{-1}(u) = \inf\{t : F_0(t) \geq u\}$ .

- b) Define the expected  $p$ -value as  $EPV(\theta) = E_\theta U$ . Let  $T_0$  denote a random variable with distribution  $F_0$ , which is independent of  $T$ . Show that  $EPV(\theta) = P(T_0 \geq T)$ .

- c) Suppose that for each  $\alpha \in (0, 1)$ , the uniformly most powerful test is of the form  $I(T \geq c)$ . Let  $EPV_T(\theta)$  be the expected  $p$ -value of  $I(T \geq c)$  and  $EPV_{T^*}(\theta)$  be the expected  $p$ -value for another test  $T^*$ . Show that for any  $\theta > 0$ ,  $EPV_T(\theta) \leq EPV_{T^*}(\theta)$ .

- d) Consider the problem of testing  $H_0 : \mu = 0$  versus  $H_1 : \mu > 0$  on the basis of  $N(\mu, 1)$  sample  $X_1, X_2, \dots, X_n$ . Let  $T = \bar{X}$ . Show that  $EPV(\theta) = \Phi(-\sqrt{n}\mu/\sqrt{2})$ , where  $\Phi$  denotes the standard normal distribution.

2. Let  $k$  and densities  $f_1, \dots, f_k$  be known, and consider an i.i.d. sample from the mixture distribution

$$f(x; \theta) = \sum_{j=1}^k \theta_j f_j(x)$$

where

$$\Theta = \{\theta \in \mathbf{R}^k : \theta_j \geq 0, \sum_{j=1}^k \theta_j = 1\}.$$

- a) Write down the equations for which the maximum likelihood estimate of  $\theta$  is the solution.



b) Describe the EM procedure for finding the MLE.

c) Calculate the information and determine the asymptotic distribution of the MLE for the (single) parameter  $\theta_1$  when  $k = 2$  and the densities  $f_1$  and  $f_2$  are variance 1 normals with unequal means.

Last Name: \_\_\_\_\_ First Name: \_\_\_\_\_

ID#: \_\_\_\_\_ Signature: \_\_\_\_\_

Do all three problems, show your partial attempts if you do not have a complete solution.

1a.) Given that  $\mathbb{P}(X \leq s, Y \leq t) = \mathbb{P}(X \leq s)\mathbb{P}(Y \leq t)$  for all real  $s, t$ , show that the random variables  $X, Y$  are independent. You may quote and use any results from measure theory that are not directly about independence.

1b.) For  $U$  chosen uniformly from  $[0, 1]$ , let  $B_i$  be the  $i$ th binary digit in the expansion of  $U$ , defined by  $B_i = 1$  if  $\lfloor 2^i U \rfloor$  is odd,  $B_i = 0$  otherwise. Show that  $B_1, B_2, B_3, \dots$  are mutually independent.

2. Let  $X_k, k \geq 1$ , be a sequence of independent normal random variables such that  $\mathbb{E}X_k = 0, k \geq 1, \mathbb{E}(X_1^2) = 1, \mathbb{E}(X_k^2) = 2^{k-2}, k \geq 2$ . Write  $S_n = X_1 + \dots + X_n, s_n^2 = \mathbb{E}S_n^2$ .

Show that the sequence does not satisfy Lindeberg's condition but CLT holds, i.e.,  $S_n/s_n$  converges in distribution to the standard normal.

(useful formula:  $\sum_{k=0}^n q^k = (q^{n+1} - 1)/(q - 1)$ ).

(The Lindeberg condition is that for all  $\varepsilon > 0$ ,

$$0 = \lim_{n \rightarrow \infty} \sum_{k=1}^n \int_{|X_k| > \varepsilon s_n} X_k^2 d\mathbb{P}.)$$

3. Assume  $X_n$  are non-negative i.i.d., and  $\mathbb{E}X_1 < \infty$ . Show that

$$\frac{X_1^2 + \dots + X_n^2}{n^2} \rightarrow 0$$

a.s., as  $n \rightarrow \infty$ .

Possible hint: for  $\varepsilon > 0$ , consider the events  $B_n = \{X_n \geq \varepsilon n\}$ , for  $n = 1, 2, \dots$

Spring 2004 Math 541b Exam

1. Ratio Estimation.

a) (Midzuno's Procedure) Let  $0 < n < N$  and  $(x_1, y_1), \dots, (x_N, y_N)$  be a fixed set of pairs of numbers with  $x_i > 0$ , and let

$$\theta = \frac{\bar{y}_N}{\bar{x}_N} = \frac{\sum_{i=1}^N y_i}{\sum_{i=1}^N x_i}.$$

Let  $I$  be a random index with distribution

$$P(I = i) = \frac{x_i}{\sum_{i=1}^N x_i}, i = 1, \dots, N,$$

and let a sample  $S$  of size  $n$  consist of  $(x_I, y_I)$  and a simple random sample of  $n - 1$  of the remaining pairs. Let

$$T = \frac{\sum_{j \in S} y_j}{\sum_{j \in S} x_j}.$$

Find  $ET$ . Hint: For a simple random sample of size  $n$ , let  $I_i$  be the indicator that pair  $i$  is included and  $\bar{x}_S$  the average of the  $x$  values in that sample. With  $I_i^*$  the indicator that pair  $i$  is included using Midzuno's scheme, show

$$E(\bar{x}_S f(I_1, \dots, I_N)) = \bar{x}_N E f(I_1^*, \dots, I_N^*).$$

b) Let  $X_i \sim \mathcal{N}(\mu_X, 1)$ ,  $Y_i \sim \mathcal{N}(\mu_Y, 1)$ ,  $i = 1, \dots, n$  be independent normal variables. Find a confidence interval for the ratio of means

$$\theta = \frac{\mu_Y}{\mu_X}$$

Hint: First consider

$$U = \bar{Y} - \theta \bar{X}.$$



2. An individual has two coins; one is unbiased and the other one is biased with head (H) probability  $p$ . The person chooses the first coin with probability  $1 - \alpha$  and the second coin with probability  $\alpha$ . Both  $p$  and  $\alpha$  are unknown parameters. He then tosses the chosen coin three times. Let  $N$  be the number of times "H" appears.

- a). What is the distribution of  $N$ ?
- b). The person does  $n$  such experiments, where in each experiment, he chooses a coin and tosses it three times. Let  $n_i$ ,  $i = 0, 1, 2, 3$  be the number of experiments in which  $i$  heads appear,  $n_0 + n_1 + n_2 + n_3 = n$ . What is the likelihood function of the observed data? What is the set of equations for the maximum likelihood estimates of  $\alpha$  and  $p$ ?
- c). Design an EM algorithm for estimating  $\alpha$  and  $p$ .
- d). How would you, in principle, use Wald's statistic to construct a  $1 - \beta$  confidence region for  $(\alpha, p)$ ? (Recall that for testing the hypothesis  $H_0 : \theta = \theta_0$  vs  $H_1 : \theta \neq \theta_0$  Wald's test statistic

$$W_n(\theta_0) = n(\hat{\theta}_n - \theta_0)^t I(\theta_0)(\hat{\theta}_n - \theta_0),$$

has an approximate  $\chi^2$ -distribution under the null hypothesis, where  $\hat{\theta}_n$  is the maximum likelihood estimate of  $\theta$  and  $I(\theta)$  is the information matrix.)

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1. Let  $X_n$  be a sequence of random variables. Show the equivalence of the following statements:

a) for each  $\epsilon > 0$ ,

$$\lim_{n \rightarrow \infty} \mathbf{P}(|X_n| > \epsilon) = 0;$$

b) For each bounded continuous function  $f$ ,

$$\mathbf{E}f(X_n) \rightarrow f(0),$$

as  $n \rightarrow \infty$ .

2. Consider a sequence of i. i. d. random variables  $X_1, X_2, \dots$ , whose probability density function is  $\pi^{-1}(1+x^2)^{-1}$  and characteristic function is  $e^{-|t|}$ .

a) What is the distribution of  $(X_1 + \dots + X_n)/n$ ?

b) Why the law of large numbers does not hold?

3.) Let  $X_1, X_2, \dots$  be i.i.d. with  $\mathbf{P}(X_i > x) = e^{-x}$  for  $x \geq 0$ .

a) Show that

$$\limsup_{n \rightarrow \infty} \frac{X_n}{\log n} = 1$$

almost surely.

b) Let  $M_n = \max_{1 \leq i \leq n} X_i$ . Show that

$$\frac{M_n}{\log n} \rightarrow 1$$

almost surely.

Math 541b Qualifying Exam (One-hour)

1. Let  $X = (X_1, \dots, X_n)$  be a sample from the uniform distribution on  $(0, \theta)$ . Show that
  - (a) For testing  $H: \theta \leq \theta_0$  against  $K: \theta > \theta_0$ , any test is UMP at level  $\alpha$  for which  $E_{\theta_0} \phi(X) = \alpha$ ,  $E_{\theta} \phi(X) \leq \alpha$  for  $\theta \leq \theta_0$ , and  $\phi(x) = 1$  when  $x_{(n)} > \theta_0$ , where we denote the order statistics by  $X_{(1)} \leq X_{(2)} \leq \dots \leq X_{(n)}$ .
  - (b) For testing  $H: \theta \leq \theta_0$  against  $K: \theta \neq \theta_0$ , a UMP test exists, and is given by  $\phi(x) = 1$  when  $x_{(n)} > \theta_0$  or  $x_{(n)} \leq \theta_0 \alpha^{1/n}$ , and by  $\phi(x) = 0$  otherwise.
2. Suppose that  $Y = (Y_1, Y_2)$ , where  $Y_1$  takes values from  $\{1, 2\}$ , and  $Y_2$  takes values from  $\{1, 2, 3\}$ . We assume that  $a_{ij} = \Pr(Y_1 = i, Y_2 = j) > 0$  for all  $(i, j)$ . We want to use Gibbs sampler to obtain the joint distribution of  $(Y_1, Y_2)$ .
  - (a) Consider the following systematic version of Gibbs sampling. In each round, we first update the value of  $Y_1$  and then the value of  $Y_2$ . Please write down the transition probability matrix for each update.
  - (b) Consider the following random-scan version of Gibbs sampling. In each step, we flip a coin with chance  $\lambda$  of obtaining a head, where  $0 < \lambda < 1$ . If it is a head, we update the value of  $Y_1$ . Otherwise, we update the value of  $Y_2$ . Please show that the associated Markov chain is in detailed balance. Show this scheme indeed converges to the joint distribution of  $(Y_1, Y_2)$ .



Math 541b Qualifying Exam, Spring 2005 (One-hour)

1. Let  $X_1, \dots, X_n$  be independent identically distributed samples from the uniform distribution  $(\theta, \theta + 1)$ ,  $\theta \in R$ . Suppose that  $n \geq 2$ . Let  $X_{(1)}, X_{(2)}, \dots, X_{(n)}$  be the order statistics from the smallest to the largest.

- (a) Show that the uniformly most powerful (UMP) test for testing  $H_0 : \theta \leq 0$  versus  $H_1 : \theta > 0$  is of the form

$$T_*(X_{(1)}, X_{(n)}) = \begin{cases} 0 & X_{(1)} < 1 - \alpha^{1/n}, X_{(n)} < 1 \\ 1 & \text{otherwise} \end{cases}$$

- (b) Find a level  $100(1 - \alpha)\%$  confidence interval for  $\theta$ .

2. Suppose that the length of life  $X$  of a light bulb manufactured by a certain process has an exponential distribution with unknown mean  $1/\theta$ , that is, the probability density function for  $X|\theta$  is

$$f(x|\theta) = \theta e^{-\theta x}.$$

Let  $X_1, X_2, \dots, X_n$  be a random sample from the population.

- (a) Prove that the gamma prior distribution for  $\theta$  with density function

$$g(\theta|\alpha, \beta) = \frac{1}{\Gamma(\alpha)\beta^\alpha} \theta^{\alpha-1} e^{-\theta/\beta}$$

is a conjugate prior.

- (b) Find the Bayesian estimate of  $\theta$  corresponding to the quadratic loss function.

3. Let  $(I_i, Y_i)$ ,  $1 \leq i \leq n$  be independent identically distributed according to  $P_\theta$ ,  $\theta = (\lambda, \mu) \in (0, 1) \times R$  where

$$P_\theta[I_1 = 1] = \lambda = 1 - P_\theta[I_1 = 0],$$

and given  $I_1 = j$ ,  $Y_1 \sim N(\mu, \sigma_j^2)$ ,  $j = 0, 1$  and  $\sigma_0 \neq \sigma_1$  known.

- (a) Find the maximum likelihood estimate of  $\theta = (\lambda, \mu)$ , when they exist.  
(b) Suppose that  $I_i, i = 1, 2, \dots, n$  are not observed. Give as explicitly as possible the E-Step and the M-step of the EM algorithm for this problem.

# **541b Qualifying Exam**

Fall, 2005

Name:

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1. Suppose  $X_1, X_2, \dots, X_n$  are independent observations from the location model with density  $f(x - \theta)$ ,  $-\infty < \theta < \infty$ , where  $f$  is differentiable and the Fisher information for  $\theta$  is finite.

a) Show that the Fisher information  $I(f)$  for  $\theta$  is constant, and compute  $I(f)$ .

We consider the test of level  $\alpha$  for  $H_n : \theta = \theta_0$  versus  $K_n : \theta = \theta_0 + h/\sqrt{n}$ , where  $h > 0$ . Under the null  $P_{\theta_0}^n$ , the following expansion is valid:

$$\log \frac{f(X_1, X_2, \dots, X_n; \theta = \theta_0 + h/\sqrt{n})}{f(X_1, X_2, \dots, X_n; \theta = \theta_0)} = \frac{h}{\sqrt{n}} \sum_{i=1}^n \frac{-f'(X_i - \theta_0)}{f(X_i - \theta_0)} - \frac{1}{2} h^2 I(f) + o_{P_{\theta_0}^n}(1).$$

b) Show that the log-likelihood-ratio tends to  $N(-\frac{1}{2}h^2 I(f), h^2 I(f))$  in distribution.

c) Show that the rejection region of the asymptotically most powerful test of level  $\alpha$  is of the form  $\sum_{i=1}^n \frac{-f'(X_i - \theta_0)}{f(X_i - \theta_0)} > c_n(\alpha)$ , for some  $c_n(\alpha)$ . Find  $c_n(\alpha)$ .

d) When  $f$  is double exponential, namely,

$$f(x - \theta) = \frac{1}{2} \exp\{-|x - \theta|\}.$$

find the asymptotically most powerful test of level  $\alpha$



2. Consider an aperiodic and irreducible Markov Chain on a finite state space  $S$  with transition matrix  $P = (p_{ij})_{i,j \in S}$ .

a) Show that if the probabilities  $\pi_i, i \in S$ , satisfy the detail balance equation

$$\pi_i p_{ij} = \pi_j p_{ji} \quad i, j \in S,$$

then they give the unique stationary distribution of the chain.

b) Let  $q_{ij}$  be a 'proposal' transition rule on  $S$ . Given  $\pi_i, i \in S$ , show how to construct transition probabilities  $p_{ij}$ , depending on  $q$  and the quantity

$$r_{ij} = \min\left\{1, \frac{\pi_j q_{ji}}{\pi_i q_{ij}}\right\},$$

which satisfy the detail balance equation. What conditions, if any, should the proposal  $q$  satisfy in order that  $\pi$  be the unique stationary distribution?

c) Let  $S_n$  be the collection of rooted binary trees on  $n$  vertices, where each vertex has either 0 or 2 descendants. Construct, in general, a Markov Chain on  $S$  that has the uniform stationary distribution, and calculate one transition probability for a simple small example. If it adds clarity, you may illustrate your proposal distribution and subsequent calculation with figures.

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1. Suppose that  $X_1, X_2, \dots$  are independent, with  $\mathbb{P}(X_n = 1) = p_n = 1 - \mathbb{P}(X_n = 0)$ .

a) Find and prove a necessary and sufficient condition, in terms of the  $p_n$ , for  $X_n \rightarrow 0$  in probability.

b) Find and prove a necessary and sufficient condition, in terms of the  $p_n$ , for  $X_n \rightarrow 0$  almost surely.

HINT: consider conditions such as  $p_n \rightarrow 0, \limsup p_n < 1, \sum_n p_n^2 < \infty, \sum_n p_n < \infty$ .

2. Suppose that  $f(x)$  is a continuous function on  $[0, 1]$ ,  $0 \leq f(x) \leq 1$ , and let  $J = \int_0^1 f(x)dx$ . Let  $(X_i, Y_i)$ ,  $i = 1, 2, \dots$  be a sequence of independent uniformly distributed over  $[0, 1]$  random variables. Let  $I_i = I_{\{f(X_i) \geq Y_i\}}$  be the indicator of the event  $\{\omega : f(X_i) \geq Y_i\}$ , and let  $J_n = n^{-1} \sum_{i=1}^n I_i$  and  $J_n^* = n^{-1} \sum_{i=1}^n f(X_i)$ ,  $n = 1, 2, \dots$

a) Why  $\lim_{n \rightarrow \infty} J_n = \lim_{n \rightarrow \infty} J_n^* = J$  with probability 1?

b) Show that the mean square error of  $J_n^*$  does not exceed the mean square error of  $J_n$  :  $E[(J_n^* - J)^2] \leq E[(J_n - J)^2]$ . For what continuous functions  $f(x)$  both errors coincide?

c) Use the CLT to find  $n$  such that  $P(|J_n - J| \leq 0.01) = 0.9$ , independently of  $f$ .

3. a) Give the definitions of the convergence in probability and convergence in distribution.

b) Let  $X$  be a Bernoulli random variable taking values 0 and 1 with equal probability  $\frac{1}{2}$ . Let  $X_1, X_2, \dots$  be identical random variables given by  $X_n = X$  for all  $n$  and let  $Y = 1 - X$ .

Does  $X_n$  converge to  $Y$  in probability? Does  $X_n$  converge to  $Y$  in distribution?

c) Prove that if a sequence of random variables  $Y_n$  converges to  $Y$  in probability, then it converges to  $X$  in distribution.



### Spring 2006 Math 541b Exam

1. Let  $X_1, \dots, X_n$  be i.i.d. samples from a Weibull distribution with density  $f(x, \lambda) = \lambda c x^{c-1} e^{-\lambda x^c}$ , where  $x > 0$ , and  $c$  is a known positive constant and  $\lambda > 0$  is the scale parameter of interest. Let  $\mu = 1/\lambda$ .
  - (a) Show that  $\sum_{i=1}^n X_i^c$  is an optimal test statistic for testing  $H: \mu = \mu_0$  versus  $K: \mu = \mu_1 > \mu_0$ . That is, the most powerful test takes the form:
 
$$\begin{cases} \text{reject } H & \text{if } \sum_{i=1}^n X_i^c > \text{critical value} \\ \text{accept } H & \text{if } \sum_{i=1}^n X_i^c \leq \text{critical value.} \end{cases}$$
  - (b) Show that  $\lambda X_i^c$  follows the standard exponential distribution  $\text{Exp}(1)$ .
  - (c) Find the critical value for the size  $\alpha$  most powerful test.
  - (d) Show that the power of the most powerful test of size  $\alpha$  is given by
 
$$\beta(\mu_1) = 1 - G_n\left(\frac{\mu_0}{\mu_1} g_n(1 - \alpha)\right),$$
 where  $G_n$  is the distribution function of  $\Gamma(n, 1)$ ,  $g_n(1 - \alpha)$  is the  $(1 - \alpha)$ th quantile of  $\Gamma(n, 1)$ , and prove that  $\beta(\mu)$  is increasing in  $\mu$ .
  - (e) Show that the most powerful test of size  $\alpha$  for the simple hypotheses in (a) is uniformly most powerful, at size  $\alpha$ , for testing the composite hypotheses  $H: \mu \leq \mu_0$  versus  $K: \mu > \mu_0$ .
  - (f) When  $n$  is large, please use normal approximation to find the critical value and power.
2. Let  $X_i, B_i, i = 1, \dots, n$  be independent Bernoulli variables where  $X_i$  has unknown success probability  $p \in (0, 1)$ , and  $B_i$  has success probability  $1/3$ . Suppose we observe
 
$$Y_i = B_i X_i + (1 - B_i)(1 - X_i), \quad i = 1, \dots, n$$
 that is, we see the original  $X_i$  with probability  $1/3$ , and  $1 - X_i$  with probability  $2/3$ .
  - (a) Write the log likelihood in term so the sum  $S_n = \sum_{i=1}^n Y_i$ , and the equation one would solve for finding the maximum likelihood estimator.

- (b) Introduce appropriate missing data for the implementation of the EM algorithm and write out the full likelihood, and the maximum likelihood estimator using this data.
- (c) Detail the steps of the EM algorithm.

Answer all three questions. Partial credit will be awarded, but in the event that you can not fully solve a problem you should state clearly what it is you have done and what you have left out. Unacknowledged omissions, incorrect reasoning and guesswork will lower your score. Start each problem on a fresh sheet of paper, and write on only one side of the paper.

1. Suppose that  $\{X_n : n \geq 1\}$  are independent identically distributed real valued random variables.

- (i) Show that  $X_n/n \rightarrow 0$  in probability.
- (ii) Show that  $X_n/n \rightarrow 0$  almost surely if and only if  $E|X_1| < \infty$ .
- (iii) Find necessary and sufficient conditions for  $X_n/\sqrt{n} \rightarrow 0$  almost surely.

2. (i) Suppose that  $X$  is an integer valued random variable with characteristic function  $\phi_X(t)$ ,  $t \in \mathbb{R}$ . Show that

$$\frac{1}{2\pi} \int_{-\pi}^{\pi} e^{-itk} \phi_X(t) dt = P(X = k)$$

for all integers  $k$ .

(ii) Now suppose that  $X_1, X_2, X_3, \dots$  are i.i.d. with the same distribution as  $X$ , and write  $S_n = X_1 + X_2 + \dots + X_n$ . Find a similar integral formula for  $P(S_n = k)$  in terms of the characteristic function  $\phi_X$ .

3. Suppose that for each  $n \geq 1$  the random variable  $X_n$  is normal with mean  $\mu_n$  and standard deviation  $\sigma_n$ .

- (i) Show that the family  $\{X_n : n \geq 1\}$  is tight if and only if  $\sup_n |\mu_n| < \infty$  and  $\sup_n \sigma_n < \infty$ .
- (ii) Show that  $X_n$  converges in distribution to some random variable  $X$  if and only if there exist  $\mu \in \mathbb{R}$  and  $\sigma \in [0, \infty)$  such that  $\mu_n \rightarrow \mu$  and  $\sigma_n \rightarrow \sigma$ .



### Spring 2007 Math 541b Exam

1. Let  $\mathbf{p} = (p_1, \dots, p_c)$  be a vector of positive numbers summing to one, and  $\mathbf{X} \sim \mathcal{M}(n, \mathbf{p})$ , the multinomial distribution given by

$$P(\mathbf{X} = \mathbf{k}) = \binom{n}{\mathbf{k}} \mathbf{p}^{\mathbf{k}},$$

where  $\mathbf{k} = (k_1, \dots, k_c)$  are non-negative integers summing to  $n$ ,

$$\binom{n}{\mathbf{k}} = \frac{n!}{\prod_{i=1}^c k_i!} \quad \text{and} \quad \mathbf{p}^{\mathbf{k}} = \prod_{i=1}^c p_i^{k_i}.$$

For a given probability vector  $\mathbf{p}_0$  we test  $H_0 : p = p_0$  versus  $H_1 : p \neq p_0$  using the chi-squared test statistic

$$V^2 = \sum_{i=1}^c \frac{(X_i - np_{i,0})^2}{np_{i,0}}.$$

- (a) Calculate the mean vector and the covariance matrix of  $\mathbf{X}$ .
- (b) Define a matrix  $\mathbf{P}$  such that

$$V^2 = n^{-1}(\mathbf{X} - n\mathbf{p})' \mathbf{P}^{-1}(\mathbf{X} - n\mathbf{p}).$$

- (c) Show that

$$n^{-1/2}(\mathbf{X} - n\mathbf{p}) \rightarrow_p Y \sim \mathcal{N}_c(0, \Sigma)$$

- (d) Find the distribution of  $U = P^{-1/2}Y$ , and show that the covariance matrix of  $U$  is a projection. (Recall that  $Q$  is a projection matrix if  $Q' = Q^2 = Q$ .) Hint: show

$$P^{-1/2}\Sigma P^{-1/2} = I - P^{-1/2}\mathbf{p}\mathbf{p}'P^{-1/2}.$$

- (e) Show that

$$V^2 \rightarrow_d \chi_{c-1}^2,$$

that is, that  $V^2$  converges in distribution to a chi squared distribution with  $c - 1$  degrees of freedom.

2. Suppose  $X_1, \dots, X_n$  are independently and identically distributed with variance  $\sigma^2$ .

- (a) Show that the estimate of variance  $\hat{\theta} = \sum_{i=1}^n (x_i - \bar{x})^2/n$  has bias equal to  $-\sigma^2/n$  as an estimator of  $\sigma^2$ .
- (b) Show that the bias of the jackknife estimate is  $-s^2/n$ , where  $s^2 = \sum_{i=1}^n (x_i - \bar{x})^2/n$ .

**MATH 507a GRADUATE EXAM**  
**FALL 2007**

Answer as many questions as you can. Partial credit will be awarded, but in the event that you can not fully solve a problem you should state clearly what it is you have done and what you have left out. Unacknowledged omissions, incorrect reasoning and guesswork will lower your score. If you cannot do part (a) of a problem, you can still get credit for (b), (c) etc. by assuming the answer to (a). Start each problem on a fresh sheet of paper, and write on only one side of the paper.

(1) In a sequence  $X_0, X_1, X_2, \dots$  of coin tosses, the length  $L_n$  of the head run starting at time  $n$  is defined by  $\{L_n \geq k\} = \{1 = X_n = X_{n+1} = \dots = X_{n+k-1}\}$ . Consider fair coin tossing, so that  $P(L_n \geq k) = 1/2^k$ . With all logs taken base 2, show that  $P(L_n > \log n + \theta \log \log n \text{ infinitely often}) = 0$  whenever  $\theta > 1$ .

(2) Suppose  $X_n, X_\infty$  are r.v.'s with characteristic functions  $\phi_n, \phi_\infty$ , all dominated by a function  $g$  in  $L^1$  (that is,  $|\phi_n(t)| \leq g(t)$  for all  $n$  and all  $t$ .) If  $\phi_n \rightarrow \phi_\infty$  pointwise, show that  $X_n$  and  $X_\infty$  have densities, call them  $f_n$  and  $f_\infty$ , and  $f_n \rightarrow f_\infty$  uniformly.

(3) Suppose  $X_n, n \geq 1$ , are r.v.'s with d.f.'s  $F_n$  satisfying  $EX_n^2 < \infty$  for all  $n$ , and

$$\lim_{A \rightarrow \infty} \sup_n \frac{\int_{\{x: |x| > A\}} x^2 dF_n(x)}{\int_{\mathbb{R}} x^2 dF_n(x)} = 0.$$

Show that  $\{F_n\}$  is tight. HINT:  $\int_{\mathbb{R}} = \int_{\{x: |x| \leq A\}} + \int_{\{x: |x| > A\}}$ .

(4)(a) Let  $\varphi \geq 0$  be a nondecreasing function on  $\mathbb{R}$ . Show that for every random variable  $Y$  and  $t \in \mathbb{R}$ ,

$$P(Y > t) \leq \frac{E\varphi(Y)}{\varphi(t)}.$$

(b) Let  $X_1, X_2, \dots$  i.i.d variables, with  $M(\lambda) := E[e^{\lambda X_1}] < \infty$  for every  $\lambda \in \mathbb{R}$ , and  $E[X_1] = 0$ . Let  $S_n = X_1 + \dots + X_n$ . Show, that for every  $x > 0$  and  $n \geq 1$

$$\frac{1}{n} \log P(S_n > nx) \leq -I(x),$$

with  $I(x) = \sup_{\lambda > 0} [\lambda x - \log M(\lambda)]$ . HINT: Use (a).



MATH 507a GRADUATE EXAM  
SPRING 2008

Answer as many questions as you can. Partial credit will be awarded, but in the event that you can not fully solve a problem you should state clearly what it is you have done and what you have left out. Unacknowledged omissions, incorrect reasoning and guesswork will lower your score. If you cannot do part (a) of a problem, you can still get credit for (b), (c) etc. by assuming the answer to (a). Start each problem on a fresh sheet of paper, and write on only one side of the paper.

(1) For  $\epsilon > 0$  let  $\{X_i^\epsilon\}$  be i.i.d. with  $P(X_i^\epsilon = \epsilon) = P(X_i^\epsilon = -\epsilon) = 1/2$ . Let  $N_\epsilon$  have a Poisson distribution with parameter  $\lambda/\epsilon^2$ , independent of the  $X_i^\epsilon$ 's. Let

$$S_\epsilon = \sum_{i=1}^{N_\epsilon} X_i^\epsilon.$$

- (a) Find the characteristic function  $\varphi_\epsilon$  of  $S_\epsilon$ .
- (b) Find  $\lim_{\epsilon \rightarrow 0} \varphi_\epsilon(t)$ . What does this tell you about the random variables  $S_\epsilon$ ?

(2) Suppose  $X_n \rightarrow X$  in distribution and  $Y_n \rightarrow 0$  in distribution. Show that  $X_n + Y_n \rightarrow X$  in distribution.

(3) Let  $U_1, U_2, \dots$  be i.i.d. sequence of Gaussian random variables with the common distribution  $\mathcal{N}(0, 1)$ . Let  $a_0, a_1, a_2, \dots$  be the real numbers such that  $a_j a_{j+1} = 0$  for all  $j \geq 0$  and that the series  $\sum a_n^2$  converge. Define

$$V_n = \sum_{k=1}^n a_{n-k} U_k, \quad n = 1, 2, \dots$$

- (a) Show that  $V_n$  and  $V_{n+1}$  are independent for all  $n \geq 1$ .
- (b) Show that with probability 1,

$$\limsup_{n \rightarrow \infty} \frac{V_n}{\sqrt{\ln n}} \leq \sqrt{2 \sum_{j=0}^{\infty} a_j^2}.$$

HINT: You can take as given the inequality  $P(U_1 \geq x) \leq \frac{1}{\sqrt{2\pi}x} e^{-x^2/2}$ , for  $x > 0$ .

(4) Prove the following inequality, sometimes known as Cantelli's inequality, and sometimes called the one-sided Chebyshev inequality: If  $X$  has mean 0 and variance 1, then for any  $c \geq 0$ ,

$$P(X \geq c) \leq \frac{1}{1 + c^2}.$$

HINT: Relate the event  $\{X \geq c\}$  to  $\{(X+t)^2 \geq (c+t)^2\}$ , for appropriate  $t$ .

Answer all three questions. Partial credit will be awarded, but in the event that you can not fully solve a problem you should state clearly what it is you have done and what you have left out. Unacknowledged omissions, incorrect reasoning and guesswork will lower your score. Start each problem on a fresh sheet of paper, and write on only one side of the paper. If you find that a calculation leads to something impossible, such as a negative probability or variance, indicate that something is wrong, but show your work anyway.

1.) Let  $\{X_n\}_{n=1}^\infty$  be a sequence of random variables that converge to  $X$  in distribution. Assume that  $P\{X_n \geq 1\} = 1$  for all  $n$ , and that  $EX_n \rightarrow c < \infty$ , as  $n \rightarrow \infty$ . Does it follow that  $E\{\ln X_n\} \rightarrow E\{\ln X\}$ ? Justify your answer.

2.) Let  $X_n, n \geq 1$  be iid Poisson random variables with parameter  $\lambda > 0$ . Show that

$$\limsup_{n \rightarrow \infty} \frac{X_n \ln(\ln n)}{\ln n} = 1$$

with probability 1.

3.) Assume that  $X, X_1, X_2, \dots$  are iid, with characteristic function  $\phi(t) = \mathbb{E} e^{itX}$ , and let  $S_n = X_1 + \dots + X_n$ .

a) For a random variable  $X$ , what special property of its characteristic function  $\phi$  holds if and only if  $X$  and  $-X$  have the same distribution? (Show both implications.)

b) Express the characteristic function of the sample average,  $\phi_{S_n/n}(t)$ , in terms of  $\phi$ .

c) Assume that  $\phi'(0) = 0$ . Show that  $S_n/n$  converges to zero in probability. (The converse is also true, but you are NOT being asked to show this.) [Hints: Show that  $\phi(u) = 1 + o(u)$  as  $u \rightarrow 0$ . Use  $\log(1+x)$  is asymptotic to  $x$  for small  $x$ , to show that for each fixed  $t$ ,  $(1+o(t/n))^n \rightarrow 1$  as  $n \rightarrow \infty$ . State how this implies the desired convergence.]

d) and e): Assume that  $X$  has density

$$f(x) = c \frac{1}{x^2 \ln|x|}$$

for  $|x| > 4$ , and  $f(x) = 0$  for  $-4 \leq x \leq 4$ , where  $c$  is the appropriate normalizing constant.

d) Show that  $\mathbb{E}|X| = \infty$ .

e) Show that the characteristic function for  $X$  has  $\phi'(0) = 0$ . [Hints: Use part a). Express  $1 - \phi(t)$  as an integral over  $x > 4$ , and use the change of variables  $y = tx$  to show that  $|1 - \phi(t)|/t \rightarrow 0$  as  $t \rightarrow 0$ . You might use  $|1 - \cos y| \leq y^2$  for all  $y$ , together with dominated convergence.]

## Fall 2008 Math 541b Exam

1. Let  $X_1, \dots, X_n$  be i.i.d. from a normal distribution with unknown mean  $\mu$  and known variance 1. Suppose that negative values of  $X_i$  are truncated at 0, so that instead of  $X_i$  we actually observe

$$Y_i = \max\{0, X_i\}, \quad i = 1, \dots, n,$$

from which we would like to estimate  $\mu$ .

- (a) Explain how to use the EM algorithm to estimate  $\mu$  from  $Y_1, \dots, Y_n$ . Specifically, give the complete log-likelihood function  $\log L_c(\mu)$  (i.e., the log of the joint density of  $X_1, \dots, X_n$ ) and a recursive formula for the successive EM estimates  $\mu^{(k+1)}$ . Write these in terms of the density  $\phi$  and c.d.f.  $\Phi$  of the standard normal distribution. *Hint:* To simplify things, assume that  $X_1, \dots, X_m$  are not truncated, and  $X_{m+1}, \dots, X_n$  are.
- (b) Find the partial log-likelihood function  $\log L(\mu)$  (i.e., the log of the joint density of  $Y_1, \dots, Y_n$ ) and use it to write down a (nonlinear) equation which the MLE  $\hat{\mu}$  satisfies. Use this equation to manually verify that  $\hat{\mu}$  is indeed a fixed point of the recursion found in (a).
2. Let  $f$  denote the true density function of  $X$ , and consider testing the simple hypotheses

$$H_0 : f = f_0 \quad \text{vs.} \quad H_1 : f = f_1$$

for given densities  $f_0, f_1$ . For a fixed value  $\pi \in (0, 1)$ , suppose that the probabilities  $\pi_0 = \pi$  and  $\pi_1 = 1 - \pi$  can be assigned to  $H_0$  and  $H_1$  prior to the experiment. We will describe tests of  $H_0$  vs.  $H_1$  by their indicator functions

$$\psi(X) = \begin{cases} 1, & \text{the test rejects } H_0 \\ 0, & \text{the test accepts } H_0. \end{cases}$$

- (a) Show that the overall probability of an error resulting from using a test  $\psi$  is

$$\pi E_0 \psi(X) + (1 - \pi) E_1 [1 - \psi(X)]. \quad (1)$$

- (b) Call the test  $\psi^*$  minimizing (1) the *Bayes optimal* test. By writing (1) as a single  $E_0$  expectation using the “change of measure” technique

$$E_1(\cdot) = E_0 \left[ (\cdot) \frac{f_1(X)}{f_0(X)} \right],$$

show that the Bayes optimal test is equivalent to a simple likelihood ratio test. Also, give the value of the likelihood ratio test’s critical value.

- (c) Argue that the Bayes optimal test is hence most powerful for detecting  $f_1$  at a certain significance level. Write down an expression for this significance level, and also give an upper bound for it as a function of  $\pi$ .
- (d) The *posterior probability* of  $H_i$  is the conditional probability that  $H_i$  is true, given  $X = x$ . Show that the posterior probability of  $H_i$  is

$$\frac{\pi_i f_i(x)}{\pi_0 f_0(x) + \pi_1 f_1(x)}. \quad (2)$$

Show that the Bayes optimal test is also equivalent to choosing which hypothesis has the larger posterior probability.



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1.) Assume  $X, X_1, X_2, \dots$  are independent identically distributed random variables, in the proper sense, having values in  $(-\infty, \infty)$ . Write  $S_n = X_1 + \dots + X_n$ . The usual SLLN (strong law of large numbers) states that if  $\mathbb{E} X = \mu \in (-\infty, \infty)$ , then  $S_n/n \rightarrow \mu$  almost surely. Use this, to prove the extended version: if  $\mathbb{E} X = \mu \in [-\infty, \infty]$ , then  $S_n/n \rightarrow \mu$  almost surely.

2.) Let  $X_n$  be a sequence of finite, independent, nonnegative random variables such that  $\lim_{n \rightarrow \infty} X_n = 0$  a.s. Prove that there is a non-random subsequence  $n_1 < n_2 < n_3 < \dots$  of the positive integers such that if

$$Y_m = X_{n_1} + X_{n_2} + \dots + X_{n_m}$$

then

$$\lim_{m \rightarrow \infty} Y_m < \infty \text{ a.s.}$$

3.) Let  $X_1, X_2, \dots$  be a sequence of random variables. For every  $n$ , the density of  $X_n$  is given by

$$f_{X_n}(x) = \frac{\text{sh}(x)}{n} \exp \left\{ \frac{1 - \text{ch}(x)}{n} \right\} \mathbf{1}(x > 0),$$

where  $\text{sh}(x) = (e^x - e^{-x})/2$  and  $\text{ch}(x) = (e^x + e^{-x})/2$ . Prove or disprove the following statements, and identify the limit in the case when you claim that the convergence holds.

a)  $\{\ln(\text{ch}(X_n)) - \ln(n)\}_{n \geq 1}$ , converges in law;

b)  $\left\{ \frac{\ln(\text{ch}(X_n))}{\ln(n)} \right\}_{n \geq 1}$ , converges in probability;

c)  $\left\{ \frac{X_n}{\ln(n)} \right\}_{n \geq 1}$ , converges in probability.

4.) Let  $X_1, X_2, \dots$  be iid, with  $E|X_1| < \infty$  finite and  $EX_1 \neq 0$ . Prove that

$$\max_{1 \leq k \leq n} \frac{|X_k|}{|S_n|} \longrightarrow 0 \text{ a.s.}$$

Hint: First show that  $\frac{|X_n|}{n} \rightarrow 0$  a.s.

Last Name: \_\_\_\_\_ First Name: \_\_\_\_\_

1. Let  $X_1, X_2, \dots$  be i.i.d. random variables uniformly distributed on  $(0, 1)$ . Prove that

$$\mathbf{P}\left\{\limsup_{n \rightarrow \infty} \left(\frac{-\log X_n}{\log n}\right) = 1\right\} = 1.$$

2. Assume  $X_1, X_2, \dots$  are independent with

$$\mathbf{P}(X_n = n^{-\alpha}) = \mathbf{P}(X_n = -n^{-\alpha}) = \frac{1}{2}.$$

For what  $\alpha$  does the series  $\sum_n X_n$  converge a.s.? For what  $\alpha$  does the series  $\sum_n |X_n|$  converge a.s.?

3. Let  $U_1, U_2, \dots$  be an i.i.d. sequence of uniform random variables on  $[0, 1]$ . Define a sequence of random variables  $\{V_n\}$  recursively as follows:

$$V_1 = U_1, \quad V_n = \begin{cases} 2V_{n-1}U_n & \text{if } V_{n-1} \in [0, \frac{1}{2}], \\ (2V_{n-1} - 1)U_n & \text{if } V_{n-1} \in [\frac{1}{2}, 1]. \end{cases}$$

- (i) Show that,  $V_{n-1}$  and  $U_n$  are independent, for all  $n \geq 2$ ;
- (ii)  $\mathbf{E}[V_n|V_{n-1}] = V_{n-1} - \frac{1}{2}\mathbf{1}_{\{1/2 \leq V_{n-1} \leq 1\}}$ , where  $\mathbf{1}_{\{1/2 \leq V_{n-1} \leq 1\}}$  is the indicator function of the set  $\{1/2 \leq V_{n-1} \leq 1\}$ ;
- (iii) Show that  $\mathbf{P}(V_{n-1} < 1/2) \rightarrow a$ , for some  $a \in [0, 1]$ , as  $n \rightarrow \infty$ . Determine the number  $a$ .

Last Name: \_\_\_\_\_ First Name: \_\_\_\_\_

1. Let  $X_1, X_2, \dots$  be a sequence of i.i.d. Poisson random variables with parameter  $\lambda > 0$ , and let  $\eta_n = \prod_{k=1}^n X_k$ .

(i) Show that  $\{\eta_n\}_{n=1}^\infty$  converges to zero in probability.

(ii) Is it possible to find a subsequence  $\{\eta_{n_k}\}_{k=1}^\infty$  and a non-zero random variable  $\eta$  with finite moment such that  $\lim_{k \rightarrow \infty} \mathbf{E}|\eta_{n_k} - \eta| = 0$ ?

2. Assume that  $X_1, X_2, \dots$  are independent random variables. Show that  $\sup_{n \geq 1} X_n < \infty$  a.s. if and only if

$$\sum_{n=1}^{\infty} \mathbf{P}(X_n > A) < \infty \text{ for some constant } A.$$

3. Let  $X_1, X_2, \dots$  be i.i.d. with  $\mathbf{E}X_i = 0$  and  $\text{Var}(X_i) = \sigma^2 > 0$ , and let  $S_n = X_1 + \dots + X_n$ . Let  $N_n$  be a sequence of integer valued random variables independent of  $X_i, i \geq 1$ , and let  $a_n$  be a sequence of positive integers with  $N_n/a_n \rightarrow 1$  in probability and  $a_n \rightarrow \infty$  as  $n \rightarrow \infty$ .

What is the limit distribution of  $\frac{S_{N_n}}{\sigma\sqrt{a_n}}$  as  $n \rightarrow \infty$ ?



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**Problem 1.** Let  $X_1, X_2, \dots$  be a sequence of (not necessarily independent and not necessarily identically distributed) real random variables defined on a common probability space. Suppose that  $\mathbb{E}(X_n^2) \leq 1$  for all  $n \geq 1$ . Does the sequence  $X_n/n$ ,  $n \geq 1$ , necessarily converge almost surely to zero? Give a proof or a counterexample.

**Problem 2.** Let  $X_1, X_2, \dots$  be iid with density

$$f(x) = \begin{cases} 0, & \text{if } |x| \leq 1 \\ |x|^{-3}, & \text{if } |x| > 1. \end{cases}$$

Prove that

$$(n \log n)^{-\frac{1}{2}} \sum_{i=1}^n X_i \xrightarrow{\mathcal{D}} \mathcal{N}(0, \sigma^2)$$

(that is, the expression on the left converges in distribution to a Gaussian random variable with mean zero and variance  $\sigma^2$ ) and determine the value of  $\sigma^2$ .

SUGGESTION: Truncate  $X_i$ ,  $i = 1, \dots, n$ , at  $\pm\sqrt{n \log n}$  and use the Central Limit Theorem for triangular arrays.

**Problem 3.** Let  $X_k$ ,  $k \geq 1$ , be iid random variables such that

$$\limsup_{n \rightarrow \infty} \frac{X_n}{n} < \infty$$

with probability one. Show that

$$\limsup_{n \rightarrow \infty} \frac{\sum_{k=1}^n X_k}{n} < \infty$$

with probability one.

SUGGESTION: Apply the Law of Large Numbers to the sequence  $\max(X_k, 0)$ ,  $k \geq 1$ .

**Problem 4.** Let  $X$  and  $Y$  be independent random variables such that  $\mathbb{E}|X + Y| < \infty$ . Is it true that  $\mathbb{E}|X| < \infty$ ? Give a proof or a counterexample.

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**Problem 1.** This is a warm-up problem on the first Borel-Cantelli lemma.

Let  $X_n$ ,  $n \geq 1$ , be independent (but not necessarily identically distributed) random variables,  $S_n = \sum_{k=1}^n X_k$ , and let  $a_n$  be real numbers such that  $a_n/a_{n+1} \leq C$  for all  $n$  and

$$P\left(\lim_{n \rightarrow \infty} \frac{S_n}{a_n} = 0\right) = 1.$$

Show that  $\sum_{k=1}^{\infty} P(|X_k| \geq a_k) < \infty$ .

**Problem 2.** This problem tests your knowledge of the basic properties of the random walk.

Let  $X_1, X_2, \dots$  be independent and identically distributed, each equal to 1 with probability  $p$  and equal to 0 with probability  $1 - p$ . Let  $S_n = \sum_{k=1}^n X_k$ .

1. Prove that if  $p \neq \frac{1}{2}$ , then, with probability 1,  $S_n = 0$  only finitely many times.
2. Prove that if  $p = \frac{1}{2}$ , then  $S_n$  will equal 0 infinitely often, but the mean recurrence time is infinite. In other words, with the notation  $\tau = \inf\{n > 1 : S_n = 0\}$ , you need to show that  $P(\tau < \infty) = 1$  but  $E\tau = +\infty$ .

**Problem 3.** This problem tests your knowledge of the strong law of large numbers.

(a) Let  $X_1, X_2, \dots$  be independent (but not necessarily identically distributed) random variables such that  $\sup_{n \geq 1} E|X_n - EX_n|^4 < \infty$ . Define  $S_n = \sum_{k=1}^n X_k$ . Give a complete proof with all the details that

$$P\left(\lim_{n \rightarrow \infty} \frac{S_n - ES_n}{n} = 0\right) = 1$$

[This result is due to Cantelli.]

(b) State, without proof, a stronger version of the result for iid random variables [due to Kolmogorov]. Please keep in mind that you cannot use this result in part (a).

**Probability (507A) Graduate Exam**  
**Fall 2011**

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1. Let  $X_n$ ,  $n \geq 1$ , and  $X$  be random variables on a probability space  $(\Omega, \mathcal{F}, \mathbb{P})$ .
  - (a) Give the definitions for “ $X_n \rightarrow X$  almost surely” and “ $X_n \rightarrow X$  in  $L^1$ ”.
  - (b) Give examples to show that (i) convergence almost surely does not imply convergence in  $L^1$ , and that (ii) convergence in  $L^1$  does not imply convergence almost surely.
  - (c) Prove that if  $\sum_{n=1}^{\infty} \mathbb{E}|X_n - X| < \infty$  then  $X_n \rightarrow X$  almost surely.
2. Let  $X_1, X_2, \dots$  be i.i.d. random variables uniformly distributed on  $(0, 1)$ . Prove that

$$P \left\{ \limsup_{n \rightarrow \infty} \left( \frac{-\log X_n}{\log n} \right) = 1 \right\} = 1.$$

3. (a) Suppose that  $X$  and  $Y$  are each uniformly distributed on  $(0, 1)$ , with  $X + Y$  constant. In the base 2 expansions of  $X$  and of  $Y$ , determine how the  $i$ th bit for  $X$  relates to the  $i$ th bit for  $Y$ .
- (b) Show that it is possible to have  $X, Y, Z$  each uniformly distributed on  $(0, 1)$ , with  $X + Y + Z$  constant. That is, give an explicit construction, or description, of the joint distribution of  $(X, Y, Z)$ . [Hint: think about the base 3 expansion of a number in  $(0, 1)$ .]