

Name \_\_\_\_\_  
Final Exam, Stat 901  
December 14, 2001. 3 Hours  
Instructor: D. L. McLeish

Do any SIX (6) of the questions below directly on the test paper.

1. Prove or disprove with a counter-example;

(a) If a sequence of events  $A_k$  satisfy  $A_k \subset A_{k+1}$  for all  $k = 1, 2, \dots$  then

$$P(\limsup_{n \rightarrow \infty} A_n) = \lim_{n \rightarrow \infty} P(A_n).$$

(b) If  $X$  has a continuous c.d.f.  $F(x)$  then the random variable  $F(X)$  has a uniform distribution on the interval  $[0, 1]$ .

(c) If  $X_i; i = 1, 2, \dots$  are random variables, then  $Y_n = \sup\{X_m; m \geq n\}$  is a random variable for each  $n$  and converges to a random variable as  $n \rightarrow \infty$ .

2. Prove or disprove with a counter-example;

(a) If  $|X|^p$  is integrable for  $p \geq 1$ , then for any constant  $\epsilon > 0$ ,

$$P[|X| \geq \epsilon] \leq \frac{E|X|^p}{\epsilon^p}$$

(b) For every integrable random variable  $X$  and every  $\epsilon > 0$ ,

$$P[|X| \geq \epsilon] < \frac{E|X|}{\epsilon}$$

(c) For any value of  $t > 0$  and random variable  $X$  with moment generating function  $m_X(t)$ ,

$$P[X > c] \leq e^{-tc} m_X(t)$$

3. Prove or disprove with a counter-example;

(a) If  $X_1, X_2, \dots, X_n$  are independent  $N(0, 1)$  random variables and  $\bar{X}_n = \frac{1}{n} \sum_{i=1}^n X_i$ , then

$$\frac{1}{n} \sum_{i=1}^n (X_i - \bar{X}_n)^2 \rightarrow 1 \text{ almost surely as } n \rightarrow \infty.$$

(b) Define  $M_n = \text{median}(X_1, X_2, \dots, X_n)$ . Then  $M_n \rightarrow 0$  almost surely as  $n \rightarrow \infty$ .

4. Prove or disprove with a counter-example;

(a) If  $F_n, F, G_n, G$  are all cumulative distribution functions and  $F_n \Rightarrow F$  and  $G_n \Rightarrow G$ , then  $F_n * G_n \Rightarrow F * G$ .

(b) If  $X_\lambda$  is a random variable with

$$P[X_\lambda = j] = \frac{\lambda^j e^{-\lambda}}{j!}, j = 0, 1, 2, \dots$$

then the characteristic function of  $X_\lambda$  is

$$\varphi(t) = \exp(\lambda(e^{it} - 1))$$

(c) As  $\lambda \rightarrow \infty$ , the distribution of

$$\frac{X_\lambda - \lambda}{\sqrt{\lambda}}$$

approaches the standard normal distribution.

5. Assume  $X$  is a random variable on the probability space  $(\Omega, \mathcal{F}, P)$  with  $E(X^2) < \infty$ . If  $\mathcal{H}$  is a sigma algebra with  $\mathcal{H} \subset \mathcal{F}$
- (a) Define  $E(X|\mathcal{H})$ .
  - (b) Prove for constants  $c, d$  that  $E(cX + d|\mathcal{H}) = cE(X|\mathcal{H}) + d$ .
  - (c) Prove if  $\mathcal{H} \subset \mathcal{G}$  are sigma-algebras,  $E[E(X|\mathcal{G})|\mathcal{H}] = E(X|\mathcal{H})$ . Does the same hold if  $\mathcal{G} \subset \mathcal{H}$ ?

6. Prove or disprove with a counter-example;

- (a) Assume  $X, Y, Z$  are random variables on the probability space  $(\Omega, \mathcal{F}, P)$ ,  $X$  is integrable and  $Y$  is independent of  $(X, Z)$ . Then.

$$E[X|Y, Z] = E[X|Z] \text{ a.s.}$$

- (b) If  $\tau$  is an optional stopping time taking values in the set  $\{1, 2, \dots, n\}$  and  $\{(X_t, \mathcal{H}_t); t = 1, 2, \dots, n\}$  is a martingale, then

$$E[(X_{j+1} - X_j)I(\tau > j)|\mathcal{H}_j] = 0 \text{ a.s. for all } j = 1, \dots, n-1.$$

- (c)  $E(X_\tau) = E(X_1)$ .

7. Prove or disprove with a counter-example ANY THREE of the following statements.

- (a) Assume the  $X_i, i = 1, 2, \dots$  are independent and identically distributed random variables with mean  $\mu$  and finite variance  $\sigma^2$ . Define  $Z_{2n} = \sum_{i=1}^n X_i - \sum_{i=n+1}^{2n} X_i$ . Then  $n^{-1/2}Z_{2n}$  converges weakly to a normal distribution..
- (b) The characteristic function  $\varphi(t)$  of every probability distribution is continuous at  $t = 0$ .
- (c) If  $X_n$  converges almost surely to a random variable  $X$  then  $X_n$  converges in probability to  $X$ .
- (d) If  $X_n$  converges weakly (in distribution) to the constant  $c$  then it converges in probability to  $c$ .