

**Qualifying Exam (August 25, 2004)**  
**Linear Algebra, Distribution Theory and Statistical**  
**Inference**

1. (a) Show that the vectors  $v^1 = \text{col}(1, 1, 0)$ ,  $v^2 = \text{col}(2, 0, 1)$ , and  $v^3 = \text{col}(1, -1, 2)$  are a basis for  $R^3$ .  
(b) Use the Gram-Schmidt procedure to construct an orthonormal basis  $\{u^1, u^2, u^3\}$  from the basis given in (a), starting with  $u^1 = v^1/\|v^1\|$  and using the standard inner product.
2. Let  $K$  and  $M$  be  $n \times n$  Hermitian matrices, and let  $M$  be positive definite. Let the vectors  $c^1, c^2, \dots, c^n$  and the real numbers  $\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_n$  satisfy

$$Kc^j = \lambda_j M c^j, \quad (M c^j, c^k) = \delta_{jk}.$$

Prove that

$$\lambda_1 = \max_{x \neq 0} \frac{(Kx, x)}{(Mx, x)},$$

where  $x$  is any  $n$ -vector. Show that the maximum is achieved when  $x = c^1$ .

3. In the following, consider  $n \times n$  matrices  $A$ .
  - (a) Show that if  $A$  has a complete set of orthogonal eigenvectors, then  $AA^* = A^*A$ .
  - (b) Use the fact that  $AA^* = A^*A$  implies that  $A$  has a complete set of orthogonal eigenvectors to prove that Hermitian, skew-Hermitian, and unitary matrices have complete sets of orthogonal eigenvectors.
  - (c) Do all matrices  $A$  with distinct eigenvalues satisfy  $AA^* = A^*A$ ? Explain.
4. Consider the random vector  $(X, Y)$ , with joint probability density function

$$f(x, y) = cx^2y, \text{ if } x^2 \leq y < 1, \\ = 0, \text{ elsewhere}$$

where  $c$  is a suitable normalizing constant. Show that  $X, Y$  are uncorrelated, but are not stochastically independent.

- 5 Let  $X_1, X_2, \dots, X_n$  be a random sample from the normal distribution with mean 0 and variance  $\sigma^2$ . Let  $\sigma_0^2 > 0$  be a given value of the population variance and a presigned  $\alpha \in (0, 1)$ . Derive a level  $\alpha$  likelihood ratio test for testing  $H_0 : \sigma^2 = \sigma_0^2$  versus the hypothesis  $H_1 : \sigma^2 \neq \sigma_0^2$  in the implementable form.
- 6 (i) Show that for any pair of random variables  $X, Y$ ,

$$\text{Var}(Y) = E\{\text{Var}(Y|X)\} + \text{Var}\{E(Y|X)\}.$$

- (ii) Draw a county in the U.S. at random. Then draw  $n$  people independently at random from the selected county. Assume  $n$  is small relative to the county population. Let  $X$  be the number of people in the sample of  $n$ , who test positive for a certain disease. If  $Q$  denotes the proportion of people who test positive for the disease; then  $Q$  is a constant (parameter) for a given county; but is a random variable if the selected county is not identified since  $Q$  may vary between counties. Based on the hierarchical model described by:  $X|\{Q = q\} \sim \text{Binomial}(n, q)$ , and  $Q \sim \text{Uniform on } (0, 1)$ ; show that the  $\text{Var}(X)$  is

$$\frac{n(n+2)}{12}.$$