## Introduction to Quantatative Methods Final Exam. September 16, 2005

1. (10%) For the matrices

$$A = \begin{bmatrix} 1 & 3 & 3 \\ 2 & 4 & 1 \end{bmatrix}, B = \begin{bmatrix} 2 & 4 \\ 1 & 5 \\ 6 & 2 \end{bmatrix}$$

compute AB, A'B' and BA.

- 2. (15%) Suppose A is a symmetric and idempotent matrix with full rank K, then
  - (a) What is the trace of *A*?
  - (b) What is the determinant of *A*?
  - (c) What is the difference between *A* and the identity matrix *I*?
- 3. (15%) Briefly answer or prove the following questions.
  - (a) (5%) Suppose A is a 2  $\times$  2 symmetric matrix, A's eigenvalues are 1, 2 and A's eigenvectors are  $c_1$  and  $c_2$ . What are the eigenvalues and eigenvectors of  $A^2$ .
  - (b) (5%) Using the fact that (AB)' = B'A' to show  $(A^{-1})' = (A')^{-1}$ .
  - (c) (5%) (True or False ?) If A is positive definite, then  $A^{-1}$  is also positive definite.
- 4. (15%) Find the vector x that minimize

$$y = x'Ax + 2a'x - 10,$$

where A is a  $K \times K$  symmetric matrix, a and x are both  $K \times 1$  vectors.

- (a) What is the first order condition for optimization? Derive the solution for x.
- (b) Suppose matrix *A* is  $\begin{bmatrix} -3 & 1 \\ 1 & -3 \end{bmatrix}$  and  $a = \begin{bmatrix} 2 \\ 2 \end{bmatrix}$ , calculate the optimal *x*.
- (c) Is the optimum in (b) a maximum or a minimum?
- 5. (15%) Let  $\{z_i\}$  be a sequence of i.i.d. (independently and identically distributed) random variables with  $E(z_i) = \mu \neq 0$  and  $Var(z_i) = \sigma^2 > 0$ , and let  $\overline{z}_n$  be the sample mean.
  - (a) (5%) What is the limiting distribution of  $\sqrt{n}(\bar{z}_n \mu)$ ?

- (b) (10%) What is the limiting distribution of  $\sqrt{n}(\sqrt{\overline{z_n}} \sqrt{\mu})$ ?
- 6. (30%) We can write a regression model in the matrix form

$$y = X \quad \beta + u$$
  
$$n \times 1 \quad n \times K \quad K \times 1 \quad n \times 1$$

Suppose *u* is distributed with zero mean and covariance matrix  $\Sigma$ , i.e.  $E(uu') = \Sigma$ , and E(X'u) = 0. It can be shown that the least square estimator of  $\beta$  is  $\hat{\beta} = (X'X)^{-1}X'y$ .

- (a) (5%) Show that  $\Sigma$  is symmetric and positive definite.
- (b) (5%) What is the covariance matrix of Pu, where P is a  $n \times n$  matrix?
- (c) (5%) Find a  $n \times n$  matrix P such that the covariance matrix of Pu is an identity matrix  $I_n$ . (Hint: think about the diagonalization and spectral decomposition of  $\Sigma$ .)
- (d) (5%) Show that  $\hat{\beta}$  is consistent.
- (e) (5%) Let  $\hat{u} \equiv y X\hat{\beta} = My$ , where  $M = I X(X'X)^{-1}X'$ . Show that M is both symmetric and idempotent.
- (f) (5%) Show that  $X'\hat{u} = 0$ . In other words,  $\hat{u}$  and X are orthogonal.

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$$AB = \begin{bmatrix} 23 & 25 \\ 14 & 30 \end{bmatrix}$$

$$A'B' = \begin{bmatrix} 1 & 2 \\ 3 & 4 \\ 3 & 1 \end{bmatrix} \begin{bmatrix} 2 & 1 & 6 \\ 4 & 5 & 2 \end{bmatrix} = \begin{bmatrix} 10 & 11 & 10 \\ 22 & 23 & 26 \\ 10 & 8 & 20 \end{bmatrix}$$

$$BA = (A'B')' = \begin{bmatrix} 10 & 22 & 10 \\ 11 & 23 & 8 \\ 10 & 26 & 20 \end{bmatrix}$$

**2a** The eigenvalues of an idempotent matrix are either 1 or 0. Since A is of full rank, all the eigenvalues are 1. Therefore, trace $(A) = \sum_{i=1}^{K} \lambda_i = K$ .

**2b**  $|A| = \prod_{i=1}^{K} \lambda_i = 1.$ 

**2c** Since  $\Lambda = \text{diag}(1, 1, \dots, 1) = I$ , then  $A = C\Lambda C' = CC' = I$ . A - I = 0.

**3a** 1,4 and *c*<sub>1</sub>, *c*<sub>2</sub>.

**3b** Since  $AA^{-1} = I$ , then  $(AA^{-1})' = (A^{-1})'A' = I' = I$ . Therefore,  $(A')^{-1} = (A^{-1})'$ .

**3c** The eigenvalues of  $A^{-1}$  are the reciprocals of eigenvalues of A which are all positive because A is positive definite. Therefore,  $A^{-1}$  is also positive definite.

**4a** The first order condition is 2Ax + 2a = 0, and  $x = -A^{-1}a$ .

**4b** 
$$x = -\begin{bmatrix} -3 & 1 \\ 1 & -3 \end{bmatrix}^{-1} \begin{bmatrix} 2 \\ 2 \end{bmatrix} = -\frac{1}{8} \begin{bmatrix} -3 & -1 \\ -1 & -3 \end{bmatrix} \begin{bmatrix} 2 \\ 2 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}.$$

**4c** The Hessian matrix is 2A. We need to check whether A is positive or negative definite. One way to check this is to find the eigenvlues of A.

$$\begin{vmatrix} -3 - \lambda & 1 \\ 1 & -3 - \lambda \end{vmatrix} = \lambda^2 + 6\lambda + 8 = 0$$
$$\lambda_1 + \lambda_2 = -6, \ \lambda_1 \lambda_2 = 8$$

Therefore,  $\lambda_1$ ,  $\lambda_2 < 0$ . *A* is negative definite. Optimum in (b) is a maximum.

5a From the Central Limit Theorem,

$$\sqrt{n}(\bar{z}_n-\mu) \stackrel{d}{\rightarrow} N(0,\sigma^2)$$

**5b** Let  $a(x) = \sqrt{x}$ , from Delta method we have

$$\sqrt{n}(a(\bar{z}_n) - a(\mu)) \xrightarrow{d} N\left(0, \left(\frac{da(\mu)}{d\mu}\right)^2 \sigma^2\right)$$

and

$$\sqrt{n}(\sqrt{\overline{z}_n} - \sqrt{\mu}) \xrightarrow{d} N\left(0, \frac{\sigma^2}{4\mu}\right).$$

**6a**  $\Sigma = E(uu')$ , then  $\Sigma' = (E(uu'))' = E(uu') = \Sigma$ ,  $\Sigma$  is symmetric. Let *a* be a *n* × 1 column vector, then

$$a'\Sigma a = E(a'uu'a) = E(\epsilon'\epsilon), \quad \epsilon \equiv u'a$$
  
=  $E(\epsilon^2) > 0$ 

 $\Sigma$  is positive definite.

6b

$$\operatorname{Var}(Pu) = \operatorname{E}(Puu'P') = P\Sigma P'$$

**6c** Since  $\Sigma = C \Lambda C'$  where columns of *C* are eigenvectors and the diagonal elements of  $\Lambda$  are eigenvalues of  $\Sigma$ . Because  $\Sigma$  is positive definite, its eigenvalues are positive and thus  $\Lambda^{1/2}$  exists.

We want to find a P such that

$$PC\Lambda^{1/2}\Lambda^{1/2}C'P' = I_n$$

Note that C'C = I, if  $P = \underline{\Lambda^{-1/2}C'}$ , then

$$\Lambda^{-1/2} C' C \Lambda^{1/2} \Lambda^{1/2} C' C \Lambda^{-1/2} = I_n$$

6d

$$\hat{\beta} = (X'X)^{-1}X'(X\beta + u)$$

$$= \beta + \left(\frac{X'X}{n}\right)^{-1}\frac{X'u}{n}$$

$$\stackrel{p}{\rightarrow} \beta + \left(\mathbb{E}(X'X)\right)^{-1} \cdot \mathbb{E}(X'u) = \beta.$$

 $\hat{\beta}$  is consistent.

6e

$$M' = I - (X(X'X)^{-1}X')' = I - X(X'X)^{-1}X' = M, M \text{ is symmetric.}$$
  

$$MM = (I - X(X'X)^{-1}X') (I - X(X'X)^{-1}X')$$
  

$$= I - X(X'X)^{-1}X' - X(X'X)^{-1}X' + X(X'X)^{-1}X'X(X'X)^{-1}X'$$
  

$$= I - X(X'X)^{-1}X' = M, M \text{ is idempotent.}$$

6f

$$\hat{X}'u = X'My = X'(I - X(X'X)^{-1}X')y$$
  
=  $X'y - X'y = 0$