



New Mexico State University  
Klipsch School of Electrical Engineering

EE555 - Advanced Linear Systems  
Spring 2005  
Exam #2

Name: \_\_\_\_\_

Solve four problems—two from each half of exam.

Prob. 1	/	25 points
Prob. 2	/	25 points
Prob. 3	/	25 points
Prob. 4	/	25 points
Prob. 5	/	25 points
Prob. 6	/	25 points
Total	/	100 points

**Prob. 1**

(i) True or False: If  $\mathbf{B} = \mathbf{U}^* \mathbf{A} \mathbf{U}$  is normal with  $\mathbf{U}$  unitary, then  $\mathbf{A}$  must be normal. If true you must prove so, otherwise simply write false (no counter-example required).

(ii) (Text p. 70) Prove that if  $\mathbf{X} \in \mathbb{C}^{n \times m}$  and  $\mathbf{X}^* \mathbf{X} = \mathbf{I}$ , then there exists a matrix  $\mathbf{Y} \in \mathbb{C}^{n \times (n-m)}$  such that  $[\mathbf{X}, \mathbf{Y}]$  is unitary. Assume  $m < n$ .

(iii) [Text Prob. IX 23(ii) p. 142] Prove that if  $\mathbf{Y} > \mathbf{0}$  and  $\begin{bmatrix} \mathbf{Y} & \mathbf{X} \\ \mathbf{X}^* & \mathbf{0} \end{bmatrix} \geq \mathbf{0} \implies \mathbf{X} = \mathbf{0}$ .

**Prob. 1 (cont.)**

**Prob. 2**

Prove the following:

(i) (Text p. 72)  $\mathbf{A}$  Hermitian  $\iff \mathbf{A}$  normal and  $\lambda_k^* = \lambda_k$  for each  $k$ .

(ii) (Text p. 72)  $\mathbf{A}$  unitary  $\iff \mathbf{A}$  normal and  $|\lambda_k| = 1$  for each  $k$ .

(iii) Prove that if  $\mathbf{A} - \mathbf{B} \geq \mathbf{0}$ , then the diagonal elements of  $\mathbf{A} - \mathbf{B}$  must be non-negative.

**Prob. 2 (cont.)**

**Prob. 3**

Let

$$\mathbf{A} = \begin{bmatrix} 1 & 2 & 0 \\ 2 & 4 & 1 \\ 1 & 2 & 0 \end{bmatrix} \text{ and } \mathbf{y} = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}.$$

(i) Find  $\mathbf{x} \in \text{Range}(\mathbf{A})$  which is nearest  $\mathbf{y}$ .

(ii) Find  $\mathbf{x} \in \text{Null}(\mathbf{A})^\perp$  which is nearest  $\mathbf{y}$ .

**Prob. 3 (cont.)**

**Prob. 4**

Prove the following:

(i)  $\text{Null}(\mathbf{C}) \subset \text{Null}(\mathbf{B})$  if and only if  $\mathbf{B}(\mathbf{I} - \mathbf{C}^+\mathbf{C}) = \mathbf{0}$ .

(ii) Suppose that  $\mathbf{A}$  is square and  $\mathbf{C} = \mathbf{C}^*$ . Then  $\text{Null}(\mathbf{C}) \subset \text{Null}(\mathbf{B})$  if and only if

$$\begin{bmatrix} \mathbf{A} & \mathbf{B} \\ \mathbf{B}^* & \mathbf{C} \end{bmatrix} = \begin{bmatrix} \mathbf{I} & \mathbf{B}\mathbf{C}^+ \\ \mathbf{0} & \mathbf{I} \end{bmatrix} \begin{bmatrix} \mathbf{A} - \mathbf{B}\mathbf{C}^+\mathbf{B}^* & \mathbf{0} \\ \mathbf{0} & \mathbf{C} \end{bmatrix} \begin{bmatrix} \mathbf{I} & \mathbf{0} \\ (\mathbf{C}^+)^*\mathbf{B}^* & \mathbf{I} \end{bmatrix}$$

Note: The corollary to (i) is  $\text{Null}(\mathbf{C}) \subset \text{Null}(\mathbf{B})$  if and only if  $(\mathbf{I} - \mathbf{C}\mathbf{C}^+)\mathbf{B}^* = \mathbf{0}$ . You may wish to use (i) and the corollary to prove (ii).

**Prob. 4 (cont.)**

**Prob. 5**

Prove the following:

(i)  $\text{Null}(\mathbf{A}) \subset \text{Null}(\mathbf{B}^*)$  if and only if  $\mathbf{B}^*(\mathbf{I} - \mathbf{A}^+\mathbf{A}) = \mathbf{0}$ .

(ii) Assuming  $\mathbf{B}^*(\mathbf{I} - \mathbf{A}^+\mathbf{A}) = \mathbf{0}$  and  $(\mathbf{I} - \mathbf{A}\mathbf{A}^+)\mathbf{B} = \mathbf{0}$ , verify the factorization

$$\begin{bmatrix} \mathbf{A} & \mathbf{B} \\ \mathbf{B}^* & \mathbf{D} \end{bmatrix} = \begin{bmatrix} \mathbf{I} & \mathbf{0} \\ \mathbf{B}^*\mathbf{A}^+ & \mathbf{I} \end{bmatrix} \begin{bmatrix} \mathbf{A} & \mathbf{0} \\ \mathbf{0} & \mathbf{D} - \mathbf{B}^*\mathbf{A}^+\mathbf{B} \end{bmatrix} \begin{bmatrix} \mathbf{I} & \mathbf{A}^+\mathbf{B} \\ \mathbf{0} & \mathbf{I} \end{bmatrix}$$

(iii) [Text Prob. IX 15(i) p. 140] Let  $\mathbf{Q} = \mathbf{Q}^* = \begin{bmatrix} \mathbf{A} & \mathbf{B} \\ \mathbf{B}^* & \mathbf{D} \end{bmatrix}$ . Prove that

$$\mathbf{Q} \geq \mathbf{0} \iff \mathbf{A} \geq \mathbf{0} \text{ and } \mathbf{D} \geq \mathbf{B}^*\mathbf{A}^+\mathbf{B} \text{ and } \text{Null}(\mathbf{A}) \subset \text{Null}(\mathbf{B}^*).$$

Note: The corollary to (i) is  $\text{Null}(\mathbf{A}) \subset \text{Null}(\mathbf{B}^*)$  if and only if  $(\mathbf{I} - \mathbf{A}\mathbf{A}^+)\mathbf{B} = \mathbf{0}$ . You may wish to use (i), (ii), and the corollary in your proof of (iii).

**Prob. 5 (cont.)**

## Prob. 6

Consider the *weighted* least-squares (LS) problem:

- Given  $\mathbf{A} \in \mathbb{C}^{n \times m}$ ,  $\mathbf{y} \in \mathbb{C}^n$ , and the “weighting matrix”  $\mathbf{W} > \mathbf{0}$
- Minimize  $V(\mathbf{x}) = (\mathbf{y} - \mathbf{Ax})^* \mathbf{W} (\mathbf{y} - \mathbf{Ax}) = \|\mathbf{y} - \mathbf{Ax}\|_{\mathbf{W}}^2$

(i) Determine the optimal solution,  $\mathbf{x}_{\text{opt}}$  assuming  $n \geq m$  and the weighted equivalent of  $\mathbf{A}^* \mathbf{A}$  (for the conventional LS problem) is non-singular. The optimal solution is the one for which  $V(\mathbf{x}) \geq V(\mathbf{x}_{\text{opt}})$ .

(ii) Determine the optimal solution,  $\mathbf{x}_{\text{opt}}$  assuming  $n \geq m$  and the weighted equivalent of  $\mathbf{A}^* \mathbf{A}$  (for the conventional LS problem) is singular.

Note: In the conventional LS problem (overdetermined) of minimizing  $V(\mathbf{x}) = \|\mathbf{y} - \mathbf{Ax}\|^2$  we had two cases of interest: one where  $\mathbf{A}^* \mathbf{A}$  is non-singular and one where  $\mathbf{A}^* \mathbf{A}$  is singular. The latter led to the use of the pseudoinverse.

**Prob. 6 (cont.)**