

**Linear Algebra**  
**Solutions to Final Exam 2010**

1.

(a) The nullspace of  $A$  is spanned by  $\begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}$  and  $\begin{bmatrix} 3 \\ 0 \\ 1 \end{bmatrix}$ , and thus  $\dim N(A) = 2$ .

By the rank-nullity theorem,  $\text{rank}(A) = n - \dim N(A) = 3 - 2 = 1$ .

(b) Let  $\mathbf{x}_1 = \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}$ ,  $\mathbf{x}_2 = \begin{bmatrix} 3 \\ 0 \\ 1 \end{bmatrix}$ . Set  $\mathbf{u}_1 = \mathbf{x}_1 = \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}$ . Compute

$$\mathbf{u}_2 = \mathbf{x}_2 - \frac{\mathbf{u}_1^T \mathbf{x}_2}{\mathbf{u}_1^T \mathbf{u}_1} \mathbf{u}_1 = \begin{bmatrix} 3 \\ 0 \\ 1 \end{bmatrix} - \frac{-3}{2} \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix} = \frac{1}{2} \begin{bmatrix} 3 \\ 3 \\ 2 \end{bmatrix}. \text{ An orthonormal basis for the}$$

nullspace of  $A$  is given by  $\mathbf{q}_1 = \frac{\mathbf{u}_1}{\|\mathbf{u}_1\|} = \frac{1}{\sqrt{2}} \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}$ ,  $\mathbf{q}_2 = \frac{\mathbf{u}_2}{\|\mathbf{u}_2\|} = \frac{1}{\sqrt{22}} \begin{bmatrix} 3 \\ 3 \\ 2 \end{bmatrix}$ .

(c) Since  $\text{rank}(A) = 1$ , the column space of  $A$  is spanned by  $\begin{bmatrix} 1 \\ 0 \\ 1 \\ 1 \end{bmatrix}$ . The

closest vector in  $C(A)$  to  $\begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}$  is its projection on  $C(A)$ , i.e.,

$$\frac{1}{3} \begin{bmatrix} 1 \\ 0 \\ 1 \\ 1 \end{bmatrix} [1 \ 0 \ 1 \ 1] \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} = \frac{1}{3} \begin{bmatrix} 1 \\ 0 \\ 1 \\ 1 \end{bmatrix}.$$

(d) Take the transpose of  $A^T \mathbf{y} = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$  and get  $\mathbf{y}^T A = [1 \ 1 \ 1]$ . It is

equivalent to ask whether  $[1 \ 1 \ 1]$  is in the row space of  $A$ . Note that the

row space of  $A$  is the orthogonal complement of the nullspace of  $A$ . But the

inner product of  $\begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$  and  $\begin{bmatrix} 3 \\ 0 \\ 1 \end{bmatrix}$  is 4, so the equation is not solvable.

- (e) Let  $\mathbf{x}$  be the eigenvector of  $Q$  satisfying  $Q\mathbf{x} = \mathbf{0}$ . This indicates that  $\mathbf{x}$  is orthogonal to the column space of  $Q$ , which is the row space of  $A$ . Hence, the eigenspace corresponding to eigenvalue 0 is the nullspace of  $A$ , which

is spanned by  $\begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 3 \\ 0 \\ 1 \end{bmatrix}$ .

- (f) NO. Since  $Q$  has zero eigenvalue, it is singular, and thus can't be an orthogonal matrix.

2.

- (a) Note that the eigenvalues of  $2A + I$  are  $-5, -3, 3$ , and thus  $\det(2A + I) = (-5)(-3)(3) = 45$ .

- (b) Note that the eigenvalues of  $A + 2I$  are  $-1, 0, 3$ , and the corresponding

eigenvectors are the same as those of  $A$ . Therefore,  $\begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$  form a basis

for the column space of  $A$  and  $\begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}$  forms a basis for the nullspace of  $A$ .

- (c) Let  $Q = [\mathbf{q}_1 \quad \mathbf{q}_2 \quad \mathbf{q}_3]$  denote the eigenvector matrix of  $A$ . Then,  $A$  can be

orthogonally diagonalized as  $A = Q\Lambda Q^T = Q \begin{bmatrix} -3 & & \\ & -2 & \\ & & 1 \end{bmatrix} Q^T$ . Then,

$$(A^{-1})^k = Q \begin{bmatrix} \left(-\frac{1}{3}\right)^k & & \\ & \left(-\frac{1}{2}\right)^k & \\ & & 1^k \end{bmatrix} Q^T \text{ approaches to}$$

$$Q \begin{bmatrix} 0 & & \\ & 0 & \\ & & 1 \end{bmatrix} Q^T = \mathbf{q}_3 \mathbf{q}_3^T = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \begin{bmatrix} 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} \text{ as } k \rightarrow \infty.$$

(d) YES. Because the eigenvalues of  $e^A$  are  $e^{\lambda_1}, e^{\lambda_2}, e^{\lambda_3}$ ,  
 $\det(e^A) = e^{\lambda_1} e^{\lambda_2} e^{\lambda_3} = e^{\lambda_1 + \lambda_2 + \lambda_3} = e^{\text{trace}(A)}$ .

(e) Since orthogonal matrices do not change the length of  $\mathbf{x}$ ,

$$\|\mathbf{Ax}\| = \|Q\Lambda Q^T \mathbf{x}\| = \|\Lambda \mathbf{y}\|, \text{ where we have defined } \mathbf{y} = Q^T \mathbf{x}. \text{ Note also that}$$

$$\|\mathbf{y}\| = \|Q^T \mathbf{x}\| = \|\mathbf{x}\| = 1. \text{ Therefore, the maximum value of}$$

$$\|\Lambda \mathbf{y}\| = \sqrt{\lambda_1^2 y_1^2 + \lambda_2^2 y_2^2 + \lambda_3^2 y_3^2} \text{ subject to unit vector } \mathbf{y} \text{ (} y_1^2 + y_2^2 + y_3^2 = 1 \text{) is}$$

determined by the maximum absolute eigenvalue of  $A$ , i.e., 3.

(f)  $\mathbf{x}^T \mathbf{Ax} = \mathbf{x}^T Q\Lambda Q^T \mathbf{x} = \mathbf{y}^T \Lambda \mathbf{y} = -3y_1^2 - 2y_2^2 + y_3^2 \geq -3$ .

(g)  $A$  is nonsingular since all its eigenvalues are nonzero.  $A$  is also symmetric since it takes the form  $A = Q\Lambda Q^T$ .  $A$  is not orthogonal since  $A^T A \neq I$ .  $A$  is diagonalizable since it has independent eigenvectors.  $A$  is not positive definite since it has negative eigenvalues.

3.

(a) FALSE. For example,  $A = \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix}$ .  $A^k$  and  $A$  have different eigenvalues

for  $k > 1$ .

(b) TRUE. The exponential  $e^A$  is symmetric alright, and has positive eigenvalues since  $e^\lambda > 0$ , where  $\lambda$  is the real eigenvalue of  $A$ .

(c) FALSE. For example,  $A = \begin{bmatrix} 1 & 0 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 2 \\ 2 & 5 \end{bmatrix}$ . The

eigenvalues of  $A$  are not 1, 1.

(d) FALSE.  $A$  is positive definite if and only if all the upper-left submatrices

have positive determinants. Here is a counterexample  $A = \begin{bmatrix} -1 & 1 \\ 1 & -2 \end{bmatrix}$ .

- (e) FALSE. Write  $\det A = \det(-A^T) = (-1)^n \det A$ . If  $n$  is an even number,  $A$  may not be singular, for example,  $A = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$ .

4.

- (a) We require that  $a > 0$  and  $\det A = a^2 - 1 > 0$ . Thus,  $a > 1$ .  
 (b) It is required that  $\operatorname{Re}(\lambda) < 0$ . Since the eigenvalues of  $A$  are  $a+1, a-1$ , we must have  $a+1 < 0$  and  $a-1 < 0$ . Hence,  $a < -1$ .  
 (c) Clearly, the eigenvalues of  $A$  must be  $\pm 1$ , and thus  $a = 0$ .

5.

- (a) Solving  $\begin{bmatrix} 1 & -1 \\ -1 & 2 \end{bmatrix} X = \begin{bmatrix} 2 & 1 \\ 1 & 1 \end{bmatrix}$  yields  $X = \begin{bmatrix} 5 & 3 \\ 3 & 2 \end{bmatrix}$ , indicating

$$[\mathbf{v}_1]_w = \begin{bmatrix} 5 \\ 3 \end{bmatrix}, [\mathbf{v}_2]_w = \begin{bmatrix} 3 \\ 2 \end{bmatrix}.$$

- (b) Let  $V = [\mathbf{v}_1 \quad \mathbf{v}_2]$ . Then,  $A = VB V^{-1}$ . Clearly,  $B = \begin{bmatrix} -2 & 0 \\ 0 & 3 \end{bmatrix}$ .

- (c) Let  $W = [\mathbf{w}_1 \quad \mathbf{w}_2]$ . We have  $A = VB V^{-1}$  and  $A = WC W^{-1}$ . Write

$$B = V^{-1} A V = V^{-1} W C W^{-1} V. \text{ Hence, } D = V^{-1} W = \begin{bmatrix} 2 & -3 \\ -3 & 5 \end{bmatrix}.$$