

## Solutions to Self-Quiz 3

- a. True. This is just part of Theorem 2 in Section 7.1. The proof appears just before the statement of the theorem.
- b. False. A counterexample is  $A = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$ .
- c. True. This is proved in the first part of the proof of Theorem 6 in Section 7.3. It is also a consequence of Theorem 7 in Section 6.2.
- d. False. The principal axes of  $\mathbf{x}^T A \mathbf{x}$  are the columns of any *orthogonal* matrix  $P$  that diagonalizes  $A$ . *Note:* When  $A$  has an eigenvalue whose eigenspace has dimension greater than 1, the principal axes are not uniquely determined.
- e. False. A counterexample is  $P = \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix}$ . The columns here are orthogonal but not orthonormal.
- f. False. See Example 6 in Section 7.2.
- g. False. A counterexample is  $A = \begin{bmatrix} 2 & 0 \\ 0 & -3 \end{bmatrix}$  and  $\mathbf{x} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$ . Then  $\mathbf{x}^T A \mathbf{x} = 2 > 0$ , but  $\mathbf{x}^T A \mathbf{x}$  is an indefinite quadratic form.
- h. True. This is basically the Principal Axes Theorem from Section 7.2. Any quadratic form can be written as  $\mathbf{x}^T A \mathbf{x}$  for some symmetric matrix  $A$ .
- i. False. See Example 3 in Section 7.3.
- j. False. The maximum value must be computed over the set of *unit* vectors. Without a restriction on the norm of  $\mathbf{x}$ , the values of  $\mathbf{x}^T A \mathbf{x}$  can be made as large as desired.
- k. False. Any orthogonal change of variable  $\mathbf{x} = P\mathbf{y}$  changes a positive definite quadratic form into another positive definite quadratic form. Proof: By Theorem 5 of Section 7.2., the classification of a quadratic form is determined by the eigenvalues of the matrix of the form. Given a form  $\mathbf{x}^T A \mathbf{x}$ , the matrix of the new quadratic form is  $P^{-1}AP$ , which is similar to  $A$  and thus has the same eigenvalues as  $A$ .
- l. False. The term “definite eigenvalue” is undefined and therefore meaningless.
- m. True. If  $\mathbf{x} = P\mathbf{y}$ , then  $\mathbf{x}^T A \mathbf{x} = (P\mathbf{y})^T A(P\mathbf{y}) = \mathbf{y}^T P^T A P \mathbf{y} = \mathbf{y}^T P^{-1} A P \mathbf{y}$ .
- n. False. A counterexample is  $U = \begin{bmatrix} 1 & -1 \\ 1 & -1 \end{bmatrix}$ . The columns of  $U$  must be *orthonormal* to make  $UU^T \mathbf{x}$  the orthogonal projection of  $\mathbf{x}$  onto  $\text{Col } U$ .
- o. True. This follows from the discussion in Example 2 of Section 7.4., which refers to a proof given in Example 1.
- p. True. Theorem 10 in Section 7.4 writes the decomposition in the form  $U \Sigma V^T$ , where  $U$  and  $V$  are orthogonal matrices. In this case,  $V^T$  is also an orthogonal matrix. Proof: Since  $V$  is orthogonal,  $V$  is invertible and  $V^{-1} = V^T$ . Then  $(V^T)^{-1} = (V^{-1})^T = (V^T)^T$ , and since  $V$  is square and invertible,  $V^T$  is an orthogonal matrix.
- q. False. A counterexample is  $A = \begin{bmatrix} 2 & 0 \\ 0 & 1 \end{bmatrix}$ . The singular values of  $A$  are 2 and 1, but the singular values of  $A^T A$  are 4 and 1.