

**THIS PAPER MUST NOT BE
REMOVED FROM THE
EXAMINATION ROOM**

**STUDENT NAME:
STUDENT NUMBER:**

Internal Students Only

THE UNIVERSITY OF QUEENSLAND

**School of Information Technology
& Electrical Engineering**

First Class Test, August 2011

ENGG7302

ADVANCED COMPUTATIONAL TECHNIQUES IN ENGINEERING

(M.E.)

CLOSED BOOK

TIME: FORTY minutes for working

FIVE minutes for perusal before examination begins

ANSWER ALL QUESTIONS ON SHEET PROVIDED

QUESTIONS CARRY THE NUMBER OF MARKS INDICATED

Drawing instruments and one battery-operated or solar-powered electronic calculator may be used but NO pre-programmed material or calculator instruction booklets are allowed in the examination room.

Part A. (1 mark each)

1. Consider a matrix $\mathbf{A} \in \mathbb{C}^{m \times n}$. Then it is in general incorrect to say that
 - (a) \mathbf{A} is invertible
 - (b) The range of \mathbf{A} is the column space of \mathbf{A}
 - (c) The rank of \mathbf{A} is $\min(m, n)$
 - (d) The nullspace of \mathbf{A} is the set of vectors that \mathbf{A} maps into $\mathbf{0}$
2. Let $S = \{\mathbf{v}_1, \mathbf{v}_2\}$, where $\mathbf{v}_1^H = (\frac{3}{5}, \frac{4}{5})$ and $\mathbf{v}_2^H = (-\frac{4}{5}, \frac{3}{5})$. Then it is true that
 - (a) S is orthogonal
 - (b) \mathbf{v}_1 and \mathbf{v}_2 are linearly independent
 - (c) S is orthonormal
 - (d) All of the above
3. Let $\mathbf{v}^H = (1, 0, 0)$ and $S = \{\mathbf{v}_1, \mathbf{v}_2\}$, where $\mathbf{q}_1^H = (1, 0, 0)$, $\mathbf{q}_2^H = (1, 0, 0)$. The component vector of \mathbf{v} orthogonal to S is
 - (a) $(0, 0, -2)$
 - (b) $(0, 0, -1)$
 - (c) $(0, 0, 2)$
 - (d) $(0, 0, 1)$

4. Given a matrix

$$\mathbf{A} = \begin{pmatrix} 2 & 1 & 1 \\ 1 & 3 & 2 \\ 1 & 1 & 3 \end{pmatrix}$$

the Frobenius norm, $\|\mathbf{A}\|_F$ is

- (a) 3
- (b) $\sqrt{31}$
- (c) 6
- (d) $\sqrt{22}$

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5. Consider replacing a matrix, \mathbf{A} , with an approximation, \mathbf{A}_ν by calculating and truncating its singular value decomposition. In what sense is \mathbf{A}_ν the best possible approximation of \mathbf{A} , compared to all other matrices of the same dimensions and rank?

- (a) \mathbf{A}_ν is an orthogonal projection matrix
- (b) \mathbf{A}_ν requires the minimum storage size
- (c) \mathbf{A}_ν can be determined using the pseudoinverse
- (d) \mathbf{A}_ν is the matrix that minimizes the 2-norm between itself and \mathbf{A}

6. If

$$\mathbf{P} = \begin{pmatrix} 0.64 & 0.48 \\ 0.48 & 0.36 \end{pmatrix}$$

is a projection matrix and $\mathbf{v} = (-3, -4)$, then a vector in the nullspace of \mathbf{P} is

- (a) $\mathbf{w}^H = (0.84, -1.12)$
- (b) $\mathbf{w}^H = (1, -1)$
- (c) $\mathbf{w}^H = (0.64, -0.3)$
- (d) $\mathbf{w}^H = (-0.64, -1)$

7. In the QR decomposition, R is matrix that is

- (a) Unitary
- (b) Diagonal
- (c) Upper triangular
- (d) Lower triangular

8. The pseudoinverse of a matrix \mathbf{A} is given by

- (a) $\mathbf{A}^+ = (\mathbf{A}^H \mathbf{A})^{-1} \mathbf{A}^H$
- (b) $\mathbf{A}^+ = (\mathbf{A}^{-1} \mathbf{A})^H \mathbf{A}^{-1}$
- (c) $\mathbf{A}^+ = (\mathbf{A} \mathbf{A})^{-1} \mathbf{A}$
- (d) $\mathbf{A}^+ = (\mathbf{A} \mathbf{A})^H \mathbf{A}$

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9. Consider the matrix

$$\mathbf{A} = \begin{pmatrix} 1 & 2 & -1 \\ 2 & 2 & 4 \\ 1 & 3 & -3 \end{pmatrix}$$

Then \mathbf{A}^{-1} is

(a)

$$\begin{pmatrix} 9 & -1.5 & -5 \\ -5 & 1 & 3 \\ -2 & 0.5 & 1 \end{pmatrix}$$

(b)

$$\begin{pmatrix} 1 & 0.5 & -0.5 \\ 0.5 & 0.5 & 0.25 \\ 1 & 0.333 & -0.333 \end{pmatrix}$$

(c)

$$\begin{pmatrix} 9 & 1 & -3 \\ -5 & 2 & -1 \\ 9 & -1.5 & -5 \end{pmatrix}$$

(d)

$$\begin{pmatrix} 1 & 2 & 1 \\ 2 & 2 & 3 \\ 1 & 3 & -3 \end{pmatrix}$$

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10. Consider the matrix \mathbf{A} whose SVD consists of

$$\mathbf{V} = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}, \mathbf{\Sigma} = \begin{pmatrix} 3 & 0 \\ 0 & 2 \\ 0 & 0 \end{pmatrix}, \mathbf{U} = \begin{pmatrix} 0 & 0 & -1 \\ 1 & 0 & 0 \\ 0 & -1 & 0 \end{pmatrix},$$

then \mathbf{A} is:

(a)

$$\begin{pmatrix} 0 & 0 \\ 0 & 3 \\ -2 & 0 \end{pmatrix}$$

(b)

$$\begin{pmatrix} 0 & 0 \\ 3 & 0 \\ -2 & -2 \end{pmatrix}$$

(c)

$$\begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 0 \end{pmatrix}$$

(d)

$$\begin{pmatrix} 0 & 0 \\ 0 & 3 \\ 2 & 1 \end{pmatrix}$$

11. If we have a QR decomposition where

$$\mathbf{Q} = \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{3}} & * \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{3}} & * \\ 0 & \frac{1}{\sqrt{3}} & * \\ -\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{3}} & * \end{pmatrix}$$

then the missing column of \mathbf{Q} is

(a) $(\frac{1}{\sqrt{2}}, 0, -\frac{1}{\sqrt{2}})^H$

(b) $(\frac{1}{6}, -\frac{1}{\sqrt{3}}, \frac{1}{6})^H$

(c) $(-\frac{1}{\sqrt{2}}, 0, \frac{1}{\sqrt{2}})^H$

(d) $(\frac{1}{6}, -\frac{1}{3}, \frac{1}{6})^H$

12. Given

$$\mathbf{A} = \begin{pmatrix} 1 & 1 \\ 1 & 2 \\ 1 & 3 \end{pmatrix}$$

and

$$\mathbf{b} = \begin{pmatrix} 2 \\ 2 \\ 4 \end{pmatrix}$$

the least squares solution of the linear system $\mathbf{Ax} = \mathbf{b}$ is

- (a) $\mathbf{x}^H = (1, 2)$
- (b) $\mathbf{x}^H = (3, 1, 1)$
- (c) $\mathbf{x}^H = (\frac{2}{3}, 1)$
- (d) $\mathbf{x}^H = (\frac{2}{3}, 1, 1)$