

MIDTERM

Instructions:

- You have 75 minutes
- This exam is open notes, books
- No computers, calculators, phones, etc
- There are 3 questions for a total of 45 points and 1 bonus question with 5 bonus points
- Good luck!

Questions:

1. [20 points] Consider the matrix and vectors

$$A = \begin{bmatrix} 0 & 4 & 1 \\ 2 & 0 & -1 \\ 0 & 1 & 0 \end{bmatrix}, \quad b = \begin{pmatrix} 2 \\ 0 \\ 1 \end{pmatrix}, \quad x = \begin{pmatrix} -1 \\ 1 \\ -2 \end{pmatrix}.$$

Answer the following questions:

- a) [2 points] Verify that x solves the equation $Ax = b$.

Multiply to verify that $Ax = b$.	(+2 points)
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- b) [2 points] Compute the LU Decomposition $A = LU$ or explain why one does not exist.

The LU Decomposition does not exist	(+1 point)
Because the first pivot, the a_{11} entry is zero.	(+1 point)

- c) [8 points] Compute the LU Decomposition $PA = LU$ with partial pivoting.
Use the computed decomposition to solve the system of equations $Ax = b$.

First iteration is:

$$P_1 = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}, \quad P_1 A_1 = \begin{bmatrix} 2 & 0 & -1 \\ 0 & 4 & 1 \\ 0 & 1 & 0 \end{bmatrix}, \quad M_1 = I, \quad M_1 P_1 A_1 = \begin{bmatrix} 2 & 0 & -1 \\ 0 & 4 & 1 \\ 0 & 1 & 0 \end{bmatrix} \quad (+1.5 \text{ points})$$

Second iteration is:

$$P_2 = I, \quad M_2 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & -1/4 & 1 \end{bmatrix}, \quad U = M_2 P_2 M_1 P_1 A_1 = \begin{bmatrix} 2 & 0 & -1 \\ 0 & 4 & 1 \\ 0 & 0 & -1/4 \end{bmatrix} \quad (+1.5 \text{ points})$$

Extracting the factors

$$P = P_2 P_1 = P_1 = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}, \quad L = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 1/4 & 1 \end{bmatrix}, \quad U = \begin{bmatrix} 2 & 0 & -1 \\ 0 & 4 & 1 \\ 0 & 0 & -1/4 \end{bmatrix} \quad (+1 \text{ point})$$

To solve $Ax = b$ we note that $PAx = LUx = Pb$ and solve

$$Ly = Pb \qquad Ux = y$$

Computing y

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 1/4 & 1 \end{bmatrix} \begin{pmatrix} y_1 \\ y_2 \\ y_3 \end{pmatrix} = Ly = Pb = \begin{pmatrix} 0 \\ 2 \\ 1 \end{pmatrix}, \quad (+1 \text{ point})$$

$$y_1 = 0, \quad y_2 = 2, \quad y_3 = 1 - (1/4)y_2 = 1 - 1/2 = 1/2 \quad (+1 \text{ point})$$

then computing x

$$Ux = \begin{bmatrix} 2 & 0 & -1 \\ 0 & 4 & 1 \\ 0 & 0 & -1/4 \end{bmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 2 \\ 1/2 \end{pmatrix} \quad (+1 \text{ point})$$

$$x_3 = -4(1/2) = -2, \quad x_2 = (2 - x_3)/4 = (2 + 2)/4 = 1, \quad x_1 = x_3/2 = -1 \quad (+1 \text{ point})$$

- d) [8 points] Compute the LU Decomposition $PAQ = LU$ with complete pivoting.
Use the computed decomposition to solve the system of equations $Ax = b$.

First iteration is:

$$P_1 = I, \quad Q_1 = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}, \quad P_1 A_1 Q_1 = \begin{bmatrix} 4 & 0 & 1 \\ 0 & 2 & -1 \\ 1 & 0 & 0 \end{bmatrix},$$

$$M_1 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ -1/4 & 0 & 1 \end{bmatrix}, \quad M_1 P_1 A_1 Q_1 = \begin{bmatrix} 4 & 0 & 1 \\ 0 & 2 & -1 \\ 0 & 0 & -1/4 \end{bmatrix} \quad (+1.5 \text{ points})$$

Second iteration is:

$$P_2 = Q_2 = M_2 = I, \quad U = M_2 P_2 M_1 P_1 A_1 Q_1 Q_2 = \begin{bmatrix} 4 & 0 & 1 \\ 0 & 2 & -1 \\ 0 & 0 & -1/4 \end{bmatrix} \quad (+1 \text{ point})$$

Extracting the factors

$$P = P_2 P_1 = I, \quad Q = Q_1 Q_2 = Q_1 = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix},$$

$$L = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 1/4 & 0 & 1 \end{bmatrix}, \quad U = \begin{bmatrix} 4 & 0 & 1 \\ 0 & 2 & -1 \\ 0 & 0 & -1/4 \end{bmatrix} \quad (+1 \text{ point})$$

To solve $Ax = b$ we note that $PAQx = LUQx = Pb$ and solve

$$Ly = Pb \qquad Uz = y, \qquad x = Q^T z$$

Computing y

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 1/4 & 0 & 1 \end{bmatrix} \begin{pmatrix} y_1 \\ y_2 \\ y_3 \end{pmatrix} = Ly = Pb = \begin{pmatrix} 2 \\ 0 \\ 1 \end{pmatrix}, \quad (+1 \text{ point})$$

$$y_1 = 2, \quad y_2 = 0, \quad y_3 = 1 - (1/4)y_1 = 1 - 1/2 = 1/2 \quad (+1 \text{ point})$$

then computing z

$$Ux = \begin{bmatrix} 4 & 0 & 1 \\ 0 & 2 & -1 \\ 0 & 0 & -1/4 \end{bmatrix} \begin{pmatrix} z_1 \\ z_2 \\ z_3 \end{pmatrix} = \begin{pmatrix} 2 \\ 0 \\ 1/2 \end{pmatrix} \quad (+1 \text{ point})$$

$$z_3 = -4(1/2) = -2, \quad z_2 = x_3/2 = -2/2 = -1, \quad z_1 = (2 - z_3)/4 = (2 + 2)/4 = 1 \quad (+1 \text{ point})$$

and finally computing x

$$x = Q^T z = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{pmatrix} 1 \\ -1 \\ -2 \end{pmatrix} = \begin{pmatrix} -1 \\ 1 \\ -2 \end{pmatrix} \quad (+.5 \text{ point})$$

2. [10 points] Consider the matrices

$$A = \begin{bmatrix} \rho & 1 \\ 0 & 1 \end{bmatrix}, \quad B = \begin{bmatrix} \rho^{-1} & 1 \\ 0 & \rho^{-1} \end{bmatrix}.$$

Answer the following questions:

a) [5 points] Using as matrix norm the $p = \infty$ induced norm show that

$$\kappa(A) = 2 + \frac{2}{\rho}, \quad \kappa(B) = (1 + \rho)^2$$

when $0 < \rho \ll 1$.

Hint: Use the fact that $\begin{bmatrix} x & 1 \\ 0 & y \end{bmatrix}^{-1} = \begin{bmatrix} x^{-1} & -x^{-1}y^{-1} \\ 0 & y^{-1} \end{bmatrix}$. Recall that $\kappa(X) = \|X\|\|X^{-1}\|$.

Compute

$$\|A\| = \max(1 + \rho, 1) = 1 + \rho, \quad \|B\| = \max(1 + \rho^{-1}, \rho^{-1}) = 1 + \rho^{-1}. \quad (+1 \text{ point})$$

Using the formula provided

$$A^{-1} = \begin{bmatrix} \rho^{-1} & -\rho^{-1} \\ 0 & 1 \end{bmatrix}, \quad B^{-1} = \begin{bmatrix} \rho & -\rho^2 \\ 0 & \rho \end{bmatrix} \quad (+1 \text{ point})$$

and

$$\|A^{-1}\| = \max(2\rho^{-1}, 1) = 2\rho^{-1}, \quad \|B^{-1}\| = \max(\rho + \rho^2, \rho) = \rho + \rho^2. \quad (+1 \text{ point})$$

Therefore

$$\kappa(A) = \|A\|\|A^{-1}\| = 2\rho^{-1}(1 + \rho) = 2 + \frac{2}{\rho} \quad (+1 \text{ point})$$

$$\kappa(B) = \|B\|\|B^{-1}\| = (1 + \rho^{-1})(\rho + \rho^2) = (1 + \rho)^2 \quad (+1 \text{ point})$$

b) [5 points] Provide a *short* explanation about the implications of the above calculations on the accuracy of solutions to the matrix equations $Ax = b$ and $Bx = b$ computed by a backward stable linear algebra algorithm when $\rho > 0$ is small.

Hint: Recall that a backward stable algorithm is one which calculates the exact solution of the perturbed problem $(A + \Delta A)y = b + \Delta b$, where $\|\Delta A\| \leq \epsilon\|A\|$, $\|\Delta b\| \leq \epsilon\|b\|$.

Based on sensitivity analysis we know the calculated solution y will have a relative error

$$\frac{\|y - x\|}{\|x\|} = O(\epsilon\kappa(X)) \quad (+1 \text{ point})$$

where x is the solution to the linear equations $Xx = b$ when $X = A$ or $X = B$.

For $X = A$ we have

$$\frac{\|y - x\|}{\|x\|} = O(\epsilon\kappa(A)) = O(\epsilon(2 + 2/\rho)) \quad (+1 \text{ point})$$

therefore we should expect large relative errors if ρ is small since $\epsilon(2 + 2/\rho) \rightarrow \infty$ for a fixed ϵ . (+1 point)

For $X = B$ we have

$$\frac{\|y - x\|}{\|x\|} = O(\epsilon\kappa(B)) = O(\epsilon(1 + \rho)) \quad (+1 \text{ point})$$

therefore we should expect small relative errors if ρ is small since $\epsilon(1 + \rho) \rightarrow \epsilon$ for a fixed ϵ . (+1 point)

3. [15 points] The matrix

$$P(v) = I - \frac{1}{v^T v} v v^T$$

is called a *projector*.

- a) [2 points] Show that $y = P(v)x = x - \alpha v$, where $\alpha = v^T x / (v^T v)$.

$$\begin{aligned} y &= P(v)x \\ &= \left(I - \frac{1}{v^T v} v v^T \right) x \\ &= x - \frac{v^T x}{v^T v} v = x - \alpha v \end{aligned} \quad (+2 \text{ points})$$

- b) [2 points] Show that $P(v)v = 0$.

With $x = v$ we have

$$\alpha = (v^T v) / (v^T v) = 1 \quad (+1 \text{ point})$$

and therefore

$$P(v)v = v - \alpha v = 0. \quad (+1 \text{ point})$$

- c) [2 points] Show that if $y = P(v)x$ then $v \perp y$.

Calculate the inner product

$$v^T y = v^T P(v)v = v^T x - \frac{v^T x}{v^T v} v^T v = 0 \quad (+2 \text{ points})$$

- d) [2 points] Show that if $v \perp x$ then $P(v)x = x$.

When $v \perp x$ then

$$v^T x = 0 \implies \alpha = 0 \quad (+1 \text{ point})$$

so that

$$P(v)x = x - \alpha v = x \quad (+1 \text{ point})$$

- e) [2 points] Prove that $P(v)$ is not full-rank unless $v = 0$.

For any $v \neq 0$ from item b) $P(v)v = 0$ hence $P(v)$ is not full-rank. (+1 point)

With $v = 0$ then $P(v) = I$ and full-rank. (+1 point)

f) [5 points] For

$$v = \begin{pmatrix} 1 \\ 0 \end{pmatrix}, \quad x_1 = \begin{pmatrix} 1 \\ 1 \end{pmatrix}, \quad x_2 = \begin{pmatrix} 3 \\ 4 \end{pmatrix},$$

compute $P(v)$, $y_1 = P(v)x_1$, and $y_2 = P(v)x_2$. Draw a cartesian plan representing the vectors v , x_1 , x_2 , y_1 and y_2 . What is the geometric interpretation of the operation $P(v)x$?

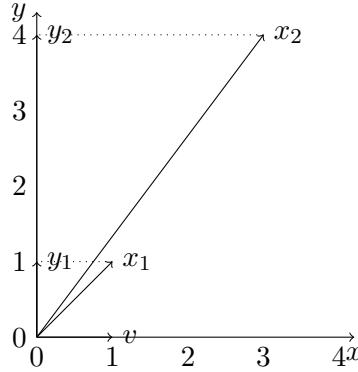
Calculating

$$P(v) = I - \frac{1}{v^T v} v v^T = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} - \begin{pmatrix} 1 \\ 0 \end{pmatrix} \begin{pmatrix} 1 & 0 \end{pmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \quad (+1 \text{ point})$$

$$y_1 = P(v)x_1 = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \begin{pmatrix} 1 \\ 1 \end{pmatrix} = \begin{pmatrix} 0 \\ 1 \end{pmatrix} \quad (+1 \text{ point})$$

$$y_2 = P(v)x_2 = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \begin{pmatrix} 3 \\ 4 \end{pmatrix} = \begin{pmatrix} 0 \\ 4 \end{pmatrix} \quad (+1 \text{ point})$$

Graphically:



which reveals that $P(v)x$ is an orthogonal projection onto axis y . (+2 points)

4. [5 bonus points] Every complex matrix $A \in \mathbb{C}^{m \times n}$, $m \leq n$, admits the complex QR Decomposition $A = QR$ where $Q \in \mathbb{C}^{m \times n}$ is such that $Q^* Q = I$ and $R \in \mathbb{C}^{n \times n}$ is upper triangular.

a) [2 bonus points] Show that matrices

$$\hat{Q} = Q \Theta, \quad \hat{R} = \Theta^* R,$$

where $\Theta \in \mathbb{C}^{n \times n}$ is a diagonal matrix with diagonal entries equal to

$$\Theta_{ii} = e^{j\theta_i}, \quad \theta_i \in [0, 2\pi), \quad i = 1, \dots, n,$$

and $j = \sqrt{-1}$, also constitute a QR Decomposition of A .

Note that $\Theta\Theta^*$ is a diagonal $n \times n$ matrix with diagonal entries

$$\Theta_{ii}\overline{\Theta_{ii}} = e^{j\theta_i}e^{-j\theta_i} = 1, \quad i = 1, \dots, n, \quad (+1 \text{ point})$$

That is $\Theta\Theta^* = \Theta^*\Theta = I$ is unitary. Therefore

$$\hat{Q}\hat{R} = Q\Theta\Theta^*R = QR = A, \quad \hat{Q}^*\hat{Q} = \Theta^*Q^*Q\Theta = \Theta^*\Theta = I \quad (+1 \text{ point})$$

is a QR decomposition because $\hat{R} = \Theta^*R$ is still upper-triangular.

- b) [2 bonus points] Explain how item a) can be used to construct a matrix $\hat{R}$ which has real and positive diagonal entries, such as the one produced by the Gram-Schmidt Algorithm.

In general r_{ii} will be a complex number. Representing r_{ii} in polar form

$$r_{ii} = |r_{ii}|e^{j\angle r_{ii}}$$

With the choice

$$\Theta_{ii} = e^{j\angle r_{ii}} \quad (+1 \text{ point})$$

The product Θ^*R is such that the diagonal entries of $\hat{R}$ are

$$\hat{r}_{ii} = e^{-j\angle r_{ii}}r_{ii} = |r_{ii}|e^{-j\angle r_{ii}}e^{j\angle r_{ii}} = |r_{ii}| \quad (+1 \text{ point})$$

- c) [1 bonus point] What are the possible values Θ can assume when A , Q , R , $\hat{Q}$ and $\hat{R}$ are real matrices?

In order for Q and $Q\Theta$ be real we will need Θ to be real. This is only possible if

$$\Theta_{ii} = e^{j\theta_i}, \quad \theta_i = \{0, \pi\}, \quad i = 1, \dots, n, \quad (+1 \text{ point})$$

in which case Θ_{ii} is either equal to '1' or '-1'.