



The
University
Of
Sheffield.

AMA226

SCHOOL OF MATHEMATICS AND STATISTICS

Autumn Semester 2008-2009

NUMERICAL LINEAR ALGEBRA

Two hours

Marks will be awarded for your best FOUR answers

- 1 (i) (a) Write down the conditions that must be satisfied by any matrix norm, $\|A\|$, and define the *subordinate matrix norm*. (5 marks)
- (b) Compute $\|A\|_1$ and $\|A\|_\infty$ for the matrix

$$A = \begin{bmatrix} 1 & -3 & 7 & -8 \\ -4 & 2 & 9 & 14 \\ 6 & 5 & -8 & 0 \\ 13 & -6 & 13 & -2 \end{bmatrix}.$$

(4 marks)

- (ii) We wish to solve $A\mathbf{x} = \mathbf{b}$ where \mathbf{b} is known exactly and A is subject to an uncertainty δA . Thus, in effect we necessarily solve

$$(A + \delta A)(\mathbf{x} + \delta \mathbf{x}) = \mathbf{b}.$$

Define the condition number, $\mathcal{K}(A)$, and hence, assuming that $\|A^{-1}\|\|\delta A\| < 1$, prove that the relative error in \mathbf{x} satisfies

$$\frac{\|\delta \mathbf{x}\|}{\|\mathbf{x}\|} \leq \frac{\mathcal{K}(A)}{\left\{1 - \mathcal{K}(A) \frac{\|\delta A\|}{\|A\|}\right\}} \frac{\|\delta A\|}{\|A\|}.$$

(9 marks)

- (iii) We wish to solve $A\mathbf{x} = \mathbf{b}$ where

$$A \approx \begin{bmatrix} 1.0000 & 0.2500 & 0.1111 & 0.0625 \\ 0.2500 & 0.1111 & 0.0625 & 0.0400 \\ 0.1111 & 0.0625 & 0.0400 & 0.0278 \\ 0.0625 & 0.0400 & 0.0278 & 0.0204 \end{bmatrix}, \quad \mathbf{b} = \begin{bmatrix} 3.14159 \\ 3.14159 \\ 3.14159 \\ 3.14159 \end{bmatrix}.$$

Given that A is subject to an uncertainty $\|\delta A\| \approx 2 \times 10^{-5}$ and \mathbf{b} is subject to an uncertainty $\|\delta \mathbf{b}\| \approx 3 \times 10^{-6}$ and that, in this case, the relative error in \mathbf{x} satisfies

$$\frac{\|\delta \mathbf{x}\|}{\|\mathbf{x}\|} \leq \frac{\mathcal{K}(A)}{\left\{1 - \mathcal{K}(A) \frac{\|\delta A\|}{\|A\|}\right\}} \left(\frac{\|\delta A\|}{\|A\|} + \frac{\|\delta \mathbf{b}\|}{\|\mathbf{b}\|} \right),$$

then:

- (a) write *scilab* code to determine bounds on $\|\delta \mathbf{x}\|_1 / \|\mathbf{x}\|_1$; (5 marks)
- (b) comment on the usefulness of solutions computed for this system given this calculation yields $\|\delta \mathbf{x}\|_1 / \|\mathbf{x}\|_1 \leq 1.42$; (2 marks)

- 2 (i) Given $m + 1$ data points (x_j, f_j) , $j = 0, 1, \dots, m$, where the x_j values are all distinct, and a set of suitably chosen basis functions, $\phi_0(x)$, $\phi_1(x)$, $\phi_2(x)$, then derive the normal equations

$$a_0 \sum_{j=0}^m \phi_0(x_j)\phi_k(x_j) + a_1 \sum_{j=0}^m \phi_1(x_j)\phi_k(x_j) + a_2 \sum_{j=0}^m \phi_2(x_j)\phi_k(x_j) = \sum_{j=0}^m f_j\phi_k(x_j),$$

$$k = 0, 1, 2$$

for determining the best least square fit to the data of the function

$$Y(x) = a_0\phi_0(x) + a_1\phi_1(x) + a_2\phi_2(x)$$

without weights.

(8 marks)

- (ii) Hence, by choosing the basis functions $\phi_0(x) \equiv 1$, $\phi_1(x) \equiv x$ and $\phi_2(x) \equiv x^2$, write down the normal equations for the best quadratic fit, and write **scilab** code to compute this fit for the data

| | | | | | | | | |
|-------|-------|-------|-------|-------|-------|-------|-------|-------|
| x_j | 1.000 | 1.500 | 2.000 | 2.500 | 3.000 | 3.500 | 4.000 | 4.500 |
| f_j | 8.00 | 11.13 | 15.00 | 18.88 | 22.00 | 23.63 | 23.00 | 19.38 |

(10 marks)

- (iii) Given that the best quadratic fit of the previous part is given by

$$P_2(x) = -2.25x^2 + 16.38x - 7.44$$

then, by computing the residuals on the sub-range $1.5 \leq x \leq 4.0$ only, comment briefly on whether or not you think the degree of the polynomial should be increased.

(7 marks)

- 3 (i) (a) Verify, by direct expansion, that the matrix equation

$$A^T A \alpha = A^T \mathbf{f} \quad (1)$$

where

$$A = \begin{pmatrix} 1 & x_0 & x_0^2 & \dots & x_0^n \\ 1 & x_1 & x_1^2 & \dots & x_1^n \\ \vdots & \vdots & \vdots & \dots & \vdots \\ 1 & x_m & x_m^2 & \dots & x_m^n \end{pmatrix}, \quad \mathbf{f} = \begin{pmatrix} f_0 \\ f_1 \\ \vdots \\ f_m \end{pmatrix}, \quad \alpha = \begin{pmatrix} a_0 \\ a_1 \\ \vdots \\ a_n \end{pmatrix}$$

represents the normal equations arising from a least squares fit of the n -th order polynomial, $P_n(x)$, to the data (x_j, f_j) , $j = 0..m$.

(3 marks)

- (b) Verify that, if P is an orthogonal matrix, then equations (1) remain as the normal equations when the transformed residuals

$$\hat{\mathbf{r}} \equiv P\mathbf{r} = P(A\alpha - \mathbf{f})$$

are minimized

(3 marks)

- (ii) For any $(m+1) \times 1$ vector \mathbf{w} satisfying $\|\mathbf{w}\| \neq 0$, an orthogonal matrix P can be defined according to

$$P = \left(I - 2 \frac{\mathbf{w}\mathbf{w}^T}{\mathbf{w}^T \mathbf{w}} \right).$$

Suppose that \mathbf{w} is defined according to

$$\mathbf{w} = \mathbf{x} + \text{sign}(x_0) \|\mathbf{x}\|_2 \mathbf{e}_0,$$

for some arbitrary $\mathbf{x} = (x_0, x_1, \dots, x_m)^T$ and where \mathbf{e}_0 is the first column of the $(m+1) \times (m+1)$ identity matrix. Show that $\hat{\mathbf{x}}$, defined according to

$$\hat{\mathbf{x}} = P\mathbf{x},$$

satisfies

$$\hat{\mathbf{x}} = -\text{sign}(x_0) \|\mathbf{x}\|_2 \mathbf{e}_0.$$

(7 marks)

3 (continued)

- (iii) It is required that $y = a_0 + a_1x + a_2x^2$ is fitted to a certain set of data in the least-squares sense. The residuals arising from this fit are given by

$$\mathbf{r} \equiv \begin{pmatrix} 1 & 0.0 & 0.00 \\ 1 & 0.2 & 0.04 \\ 1 & 0.4 & 0.16 \\ \vdots & \vdots & \vdots \\ 1 & 2.0 & 4.00 \end{pmatrix} \begin{pmatrix} a_0 \\ a_1 \\ a_2 \end{pmatrix} - \begin{pmatrix} 0.4000 \\ 0.2046 \\ 0.0329 \\ \vdots \\ 0.5972 \end{pmatrix}$$

where (a_0, a_1, a_2) are chosen such that $\|\mathbf{r}\|_2$ is minimised. A series of orthogonal transformations $P \equiv P_4P_3P_2P_1$, based on the Householder Reflection matrix, is applied to \mathbf{r} to obtain $\hat{\mathbf{r}} \equiv P\mathbf{r}$ where

$$\hat{\mathbf{r}} \equiv \begin{pmatrix} -3.3166 & -3.3166 & -4.6433 \\ 0.0000 & 2.0977 & 4.1952 \\ 0.0000 & 0.0000 & 1.1717 \\ 0.0000 & 0.0000 & 0.0000 \\ 0.0000 & 0.0000 & 0.0000 \\ \vdots & \vdots & \vdots \\ 0.0000 & 0.0000 & 0.0000 \end{pmatrix} \begin{pmatrix} a_0 \\ a_1 \\ a_2 \end{pmatrix} - \begin{pmatrix} -0.2517 \\ 0.1432 \\ 0.8349 \\ -0.0662 \\ 0.0000 \\ \vdots \\ 0.0000 \end{pmatrix}.$$

Determine the values of (a_0, a_1, a_2) which minimizes $\|\hat{\mathbf{r}}\|_2$ and state the value of this minimized $\|\hat{\mathbf{r}}\|_2$. Work correct to four decimal places.

(7 marks)

- (iv) We wish to apply a single Householder reflection to the matrix

$$A = \begin{bmatrix} 1 & 2.2 \\ 1 & 2.4 \\ 1 & 2.6 \\ 1 & 2.8 \\ 1 & 3.0 \end{bmatrix}.$$

Write *scilab* code to accomplish this task.

(5 marks)

- 4 (i) (a) The real symmetric matrix A has eigenvalues $\lambda_1, \lambda_2, \dots, \lambda_n$ satisfying

$$|\lambda_1| > |\lambda_2| \geq |\lambda_3| \geq \dots \geq |\lambda_n| > 0$$

with corresponding linearly independent eigenvectors $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n$ which can be supposed normalized so that the largest element of each one is unity. The iteration

$$\begin{aligned} \mathbf{y}_k &= A\mathbf{z}_{k-1} \\ \mathbf{z}_k &= \frac{\mathbf{y}_k}{\mu_k}, \quad k = 1, 2, \dots, \end{aligned}$$

where \mathbf{z}_0 is normalized so that its largest element is unity, and where μ_k is chosen so that the largest element of \mathbf{z}_k is likewise unity, is called the *Power Method*. Prove that, for this iteration, $(\mu_k, \mathbf{z}_k) \rightarrow (\lambda_1, \mathbf{x}_1)$ as $k \rightarrow \infty$. **(10 marks)**

- (b) Supposing that A and \mathbf{z}_0 are already defined, then write **scilab** code which impliments ten iterations of the Power Method to produce an estimate of $(\lambda_1, \mathbf{x}_1)$. **(5 marks)**

- (ii) (a) Suppose now that A is a real symmetric matrix and that an estimate of $(\lambda_2, \mathbf{x}_2)$ is also required. The method of *Hotelling's Deflation*,

$$B = A - \lambda_1 \mathbf{x}_1 \mathbf{x}_1^T,$$

where the eigenvectors are supposed normalized so that $\mathbf{x}_k^T \mathbf{x}_k = 1$, is useful for this purpose since B has eigenvalues $0, \lambda_2, \lambda_3, \dots, \lambda_n$. Prove this latter statement. **(5 marks)**

- (b) Write additional **scilab** code which takes the results of your Power Method code (above) and uses Hotelling's Deflation to produce the matrix B , and then applies ten iterations of the Power Method to the matrix B to obtain an estimate of $(\lambda_2, \mathbf{x}_2)$. **(5 marks)**

- 5 (i) The linear system

$$Ax = \mathbf{b},$$

where A is an $n \times n$ matrix of known coefficients, \mathbf{b} is an $n \times 1$ column vector of known values can be rearranged, for the solution vector \mathbf{x} , in arbitrarily many ways in the form

$$\mathbf{x} = H\mathbf{x} + \mathbf{d}$$

which can subsequently be used to define the iteration

$$\mathbf{x}^{(k+1)} = H\mathbf{x}^{(k)} + \mathbf{d} \quad (2)$$

where H is some $n \times n$ matrix and \mathbf{d} is a column vector of known values.

- (a) Derive a sufficient condition, written in terms of $\|H\|$, which will guarantee that the iteration (2) above will give a sequence of iterates convergent to \mathbf{x} , the solution of $Ax = \mathbf{b}$. **(5 marks)**
- (b) Starting from $Ax = \mathbf{b}$, write down the Jacobi iteration, and hence prove that **strict diagonal dominance** of the matrix A is sufficient to guarantee the convergence of the method. **(8 marks)**
- (ii) (a) Use the Jacobi iterative method to obtain two successive approximations to the solution of the system $Ax = \mathbf{b}$ where

$$A = \begin{pmatrix} 12 & -5 & -2 \\ -5 & 12 & -3 \\ -2 & -3 & 12 \end{pmatrix} \quad \text{and} \quad \mathbf{b} = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix},$$

using $\mathbf{x}^{(0)} = (0, 0, 0)^T$ as the starting vector and working correct to four significant figures. **(5 marks)**

- (b) Write **scilab** code which implements the Jacobi method to generate an approximate solution to the system $Ax = \mathbf{b}$ defined above that will iterate until $\|Ax - \mathbf{b}\|_{\infty} \leq 10^{-5}$. **(7 marks)**

End of Question Paper