

# UNIVERSITY OF OSLO

Faculty of mathematics and natural sciences

Examination in INF-MAT 4350 — Numerical linear algebra

Day of examination: 7 December 2012

Examination hours: 0900–1300

This problem set consists of 4 pages.

Appendices: None

Permitted aids: None

Please make sure that your copy of the problem set is complete before you attempt to answer anything.

All 9 part questions will be weighted equally.

## Problem 1 Gauss-Seidel

Consider the matrix

$$\mathbf{A} := \begin{bmatrix} 4 & -\alpha \\ -\alpha & 1 \end{bmatrix}, \quad \alpha \in \mathbb{R}.$$

### 1a

For what values of  $\alpha$  is  $\mathbf{A}$  symmetric positive definite?

**Answer:**  $\mathbf{A}$  is symmetric for any  $\alpha$ . Since  $a_{11} > 0$ , the matrix  $\mathbf{A}$  is positive definite if and only if  $\det(\mathbf{A}) = 4 - \alpha^2 > 0$  or  $-2 < \alpha < 2$ .

### 1b

For what values of  $\alpha$  does Gauss Seidel's method converge?

**Answer:** Applying GS to the system

$$\begin{bmatrix} 4 & -\alpha \\ -\alpha & 1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} b \\ c \end{bmatrix}$$

we find  $x_{k+1} = \frac{\alpha}{4}y_k + \frac{1}{4}b$  and  $y_{k+1} = \alpha x_{k+1} + c = \alpha(\frac{\alpha}{4}y_k + \frac{1}{4}b) + c$ . Thus

$$\begin{bmatrix} x_{k+1} \\ y_{k+1} \end{bmatrix} = \mathbf{G} \begin{bmatrix} x_k \\ y_k \end{bmatrix} + \mathbf{c}, \quad \mathbf{G} = \begin{bmatrix} 0 & \alpha/4 \\ 0 & \alpha^2/4 \end{bmatrix}.$$

GS converges if and only if  $\rho(\mathbf{G}) < 1$ . Since  $\mathbf{G}$  has eigenvalues 0 and  $\alpha^2/4$  this happens if and only if  $-2 < \alpha < 2$  i.e., if and only if  $\mathbf{A}$  is positive definite.

(Continued on page 2.)

## Problem 2 Perturbation

Let  $\|\cdot\|$  be a vector norm on  $\mathbb{R}^n$  and for any  $\mathbf{B} \in \mathbb{R}^{n \times n}$  let

$$\|\mathbf{B}\| := \max_{\mathbf{x} \neq \mathbf{0}} \frac{\|\mathbf{B}\mathbf{x}\|}{\|\mathbf{x}\|}$$

be the associated operator norm of  $\mathbf{B}$ . Suppose  $\mathbf{A} \in \mathbb{R}^{n \times n}$  is nonsingular.

### 2a

Show that for any  $\mathbf{b}, \mathbf{e} \in \mathbb{R}^n$  with  $\mathbf{b} \neq \mathbf{0}$

$$\frac{\|\mathbf{e}\|}{\|\mathbf{b}\|} \leq \|\mathbf{A}\| \|\mathbf{A}^{-1}\| \frac{\|\mathbf{y} - \mathbf{x}\|}{\|\mathbf{x}\|}, \quad (1)$$

where  $\mathbf{x}$  and  $\mathbf{y}$  are solutions of  $\mathbf{A}\mathbf{x} = \mathbf{b}$  and  $\mathbf{A}\mathbf{y} = \mathbf{b} + \mathbf{e}$ .

Hint: Use that  $\mathbf{A}(\mathbf{y} - \mathbf{x}) = \mathbf{e}$  and  $\mathbf{x} = \mathbf{A}^{-1}\mathbf{b}$ .

**Answer:** Subtracting  $\mathbf{A}\mathbf{x} = \mathbf{b}$  from  $\mathbf{A}\mathbf{y} = \mathbf{b} + \mathbf{e}$  we find  $\mathbf{A}(\mathbf{y} - \mathbf{x}) = \mathbf{e}$ . Taking norms

$$\|\mathbf{e}\| \leq \|\mathbf{A}\| \|\mathbf{y} - \mathbf{x}\|, \quad \|\mathbf{x}\| \leq \|\mathbf{A}^{-1}\| \|\mathbf{b}\|.$$

But then  $\frac{1}{\|\mathbf{b}\|} \leq \frac{\|\mathbf{A}^{-1}\|}{\|\mathbf{x}\|}$ ,

$$\frac{\|\mathbf{e}\|}{\|\mathbf{b}\|} \leq \|\mathbf{A}\| \|\mathbf{y} - \mathbf{x}\| \frac{\|\mathbf{A}^{-1}\|}{\|\mathbf{x}\|}.$$

and (1) follows.

### 2b

Show that we have equality in (1) for some vectors  $\mathbf{b}$  and  $\mathbf{e}$ .

Hint: There are vectors  $\mathbf{c}$  and  $\mathbf{d}$  so that

$$\|\mathbf{A}^{-1}\| = \frac{\|\mathbf{A}^{-1}\mathbf{c}\|}{\|\mathbf{c}\|}, \quad \|\mathbf{A}\| = \frac{\|\mathbf{A}\mathbf{d}\|}{\|\mathbf{d}\|}.$$

You should not show this.

**Answer:** Define  $\mathbf{b} := \mathbf{c}$  and  $\mathbf{e} := \mathbf{A}\mathbf{d}$ . Then  $\|\mathbf{e}\|/\|\mathbf{b}\| = \|\mathbf{A}\mathbf{d}\|/\|\mathbf{c}\|$ . Now

$$\frac{1}{\|\mathbf{c}\|} = \frac{\|\mathbf{A}^{-1}\|}{\|\mathbf{A}^{-1}\mathbf{c}\|} = \frac{\|\mathbf{A}^{-1}\|}{\|\mathbf{x}\|}, \quad \|\mathbf{A}\mathbf{d}\| = \|\mathbf{A}\| \|\mathbf{d}\| = \|\mathbf{A}\| \|\mathbf{A}^{-1}\mathbf{e}\| = \|\mathbf{A}\| \|\mathbf{y} - \mathbf{x}\|.$$

But then

$$\frac{\|\mathbf{e}\|}{\|\mathbf{b}\|} = \frac{\|\mathbf{A}\mathbf{d}\|}{\|\mathbf{c}\|} = \frac{\|\mathbf{A}^{-1}\|}{\|\mathbf{x}\|} \|\mathbf{A}\| \|\mathbf{y} - \mathbf{x}\|$$

and (1) holds with equality.

(Continued on page 3.)

### Problem 3 Eigenvalue bound

In this exercise we assume that  $\mathbf{A} \in \mathbb{R}^{n \times n}$  has eigenpairs  $(\lambda_j, \mathbf{x}_j)$ ,  $j = 1, \dots, n$ , where the eigenvector matrix  $\mathbf{X} = [\mathbf{x}_1, \dots, \mathbf{x}_n]$  is nonsingular. We know that  $\mathbf{A} = \mathbf{X}\mathbf{D}\mathbf{X}^{-1}$ , where  $\mathbf{D} = \text{diag}(\lambda_1, \dots, \lambda_n)$ . We let  $\|\mathbf{A}\|_2 := \max_{\mathbf{x} \neq \mathbf{0}} \|\mathbf{A}\mathbf{x}\|_2 / \|\mathbf{x}\|_2$  be the spectral norm of  $\mathbf{A}$ .

We want to show the following theorem:

#### Theorem 1

To any  $\mu \in \mathbb{R}$  with  $\mu - \lambda_j \neq 0$  for  $j = 1, \dots, n$ . and  $\mathbf{x} \in \mathbb{R}^n$  with  $\|\mathbf{x}\|_2 = 1$  we can find an eigenvalue  $\lambda$  of  $\mathbf{A}$  such that

$$|\lambda - \mu| \leq K_2(\mathbf{X}) \|\mathbf{r}\|_2,$$

where  $\mathbf{r} := \mathbf{A}\mathbf{x} - \mu\mathbf{x}$  and  $K_2(\mathbf{X}) := \|\mathbf{X}\|_2 \|\mathbf{X}^{-1}\|_2$ .

#### 3a

Show that  $\|\mathbf{D}\|_2 = \rho(\mathbf{A}) := \max_i |\lambda_i|$ .

**Answer:** Since  $\|\mathbf{D}\|_2$  equals the square root of the largest eigenvalue of  $\mathbf{D}^T \mathbf{D}$  we have

$$\|\mathbf{D}\|_2 = \sqrt{\rho(\mathbf{D}^T \mathbf{D})} = \sqrt{\rho(\mathbf{D}^2)} = \sqrt{\rho(\mathbf{D})^2} = \rho(\mathbf{D}) = \rho(\mathbf{A}).$$

#### 3b

We define  $\mathbf{D}_1 := \mathbf{D} - \mu\mathbf{I}$ . Show that  $\mathbf{D}_1$  is nonsingular and  $\|\mathbf{D}_1^{-1}\|_2 = \frac{1}{\lambda - \mu}$ , where  $|\lambda - \mu| := \min_j |\lambda_j - \mu|$ .

**Answer:**  $\mathbf{D}_1$  is nonsingular since it is a diagonal matrix with nonzero diagonal elements  $\lambda_j - \mu$  for  $j = 1, \dots, n$ . We have  $\mathbf{D}_1^{-1} = \text{diag}((\lambda_1 - \mu)^{-1}, \dots, (\lambda_n - \mu)^{-1})$ . The result follows from problem 3a with  $\mathbf{D} = \mathbf{D}_1$ .

#### 3c

Show that  $\mathbf{X}\mathbf{D}_1^{-1}\mathbf{X}^{-1}\mathbf{r} = \mathbf{x}$ , where  $\mathbf{r} := \mathbf{A}\mathbf{x} - \mu\mathbf{x}$ .

**Answer:**

$$\mathbf{X}\mathbf{D}_1^{-1}\mathbf{X}^{-1}\mathbf{r} = (\mathbf{X}(\mathbf{D} - \mu\mathbf{I})\mathbf{X}^{-1})^{-1}\mathbf{r} = (\mathbf{A} - \mu\mathbf{I})^{-1}(\mathbf{A} - \mu\mathbf{I})\mathbf{x} = \mathbf{x}.$$

#### 3d

Show Theorem 1.

**Answer:** Let  $|\lambda - \mu| = \min_j |\lambda_j - \mu|$ .

$$1 = \|\mathbf{x}\|_2 = \|\mathbf{X}\mathbf{D}_1^{-1}\mathbf{X}^{-1}\mathbf{r}\|_2 \leq \|\mathbf{D}_1^{-1}\|_2 K_2(\mathbf{X}) \|\mathbf{r}\|_2 = \frac{K_2(\mathbf{X}) \|\mathbf{r}\|_2}{\min_j |\lambda_j - \mu|} = \frac{K_2(\mathbf{X}) \|\mathbf{r}\|_2}{|\lambda - \mu|}.$$

But then the theorem follows.

(Continued on page 4.)

## Problem 4 Matlab program

Recall that a square matrix  $\mathbf{A}$  is  $d$ -banded if  $a_{ij} = 0$  for  $|i - j| > d$ . Write a Matlab function `x=backsolve(A,b,d)` that for a given nonsingular upper triangular  $d$ -banded matrix  $\mathbf{A} \in \mathbb{R}^{n \times n}$  and  $\mathbf{b} \in \mathbb{R}^n$  computes a solution  $\mathbf{x}$  to the system  $\mathbf{Ax} = \mathbf{b}$ .

**Answer:**

```
function x=backsolve(A,b,d)
n=length(b); x=b;
x(n)=b(n)/A(n,n);
for k=n-1:-1:1
    uk=min(n,k+d);
    x(k)=(b(k)-A(k,k+1:uk)*x(k+1:uk))/A(k,k);
end
```

Good luck!