



Department of Mathematics  
University of Houston  
Numerical Analysis I  
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## Numerical Analysis I (4th Homework Assignment)

### Exercise 13 (*Gradient method and semi-iterative Richardson iteration*)

Let  $A \in \mathbf{R}^{n \times n}$  be symmetric positive definite with the extreme eigenvalues  $\lambda := \lambda_{\min}(A)$ ,  $\Lambda := \lambda_{\max}(A)$  and let  $\kappa := \Lambda/\lambda$  be the spectral condition of  $A$ . Further, let  $b \in \mathbf{R}^n$  and  $x^0 \in \mathbf{R}^n$ .

(i) Consider the semi-iterative Richardson iteration

$$y^{m+1} = y^m + \Theta_{m+1}(b - Ay^m), \quad m \geq 0.$$

Specify the values of  $y^0 \in \mathbf{R}^n$  and  $\Theta_{m+1}$  for which the sequence of iterates  $(y^m)_{\mathbf{N}}$  corresponds to the sequence  $(x^m)_{\mathbf{N}}$  obtained by the gradient method.

(ii) Show that for any initial vector  $x^0 \in \mathbf{R}^n$  the sequence  $(x^m)_{\mathbf{N}}$ , obtained by the gradient method, converges to  $x^* := A^{-1}b$ . Moreover, verify the estimates

$$F(x^m) - F(x^*) \leq \left(\frac{\kappa - 1}{\kappa + 1}\right)^{2m} [F(x^0) - F(x^*)],$$

$$\|x^m - x^*\|_A \leq \left(\frac{\kappa - 1}{\kappa + 1}\right)^m \|x^0 - x^*\|_A,$$

where  $F(x) := \frac{1}{2} \langle Ax, x \rangle - \langle b, x \rangle$ .

[Hint: Utilize Theorem 1.9 as presented in class and take advantage of the fact that  $x^m$  minimizes the error  $\|x^m - x^*\|_A$ .]

### Exercise 14 (*Convergence of the cg method*)

Assume that  $A \in \mathbf{R}^{n \times n}$  is symmetric positive definite with  $p < n$  different eigenvalues. Consider the linear system  $Ax = b \in \mathbf{R}^n$  and show that in case of exact computations, for any initial iterate  $x^0 \in \mathbf{R}^n$  the cg method terminates after at most  $p$  steps with the solution  $A^{-1}b$ .

[Hint: Apply the Cayley-Hamilton theorem together with estimates for the iteration error.]

**Exercise 15** (*Preconditioned cg method*)

The numerical solution of two-point boundary value problems often gives rise to a symmetric matrix  $A = (a_{ij})_{i,j=1}^n \in \mathbf{R}^{n \times n}$ , where

$$a_{ij} = \begin{cases} 2 & \text{für } i = j \\ -1 & \text{für } |i - j| = 1 \\ 0 & \text{für } |i - j| \geq 2 \end{cases}, \quad 1 \leq i, j \leq n.$$

(i) Show that  $A$  is positive definite.

(ii) Let  $D := \text{diag}(A)$  be the diagonal of  $A$  and denote by  $L$  the lower triangular part so that  $A = L + D + L^T$ .

Consider the linear system  $Ax = b$ ,  $b \in \mathbf{R}^n$  and show that the preconditioned cg method with preconditioner  $C := EE^T$ , where  $E := \frac{1}{2}D + L$ , converges after at most two steps.

[Hint: Observe that the convergence properties of the preconditioned cg method correspond to the original cg method applied to the transformed matrix  $\tilde{A} := E^{-1}AE^{-T}$  and apply the result of Exercise 14.]

**Exercise 16** (*cg method for the discrete Laplacian*)

The discretization of the 2D Laplacian by finite differences with respect to a uniform grid results in a linear algebraic system with the symmetric coefficient matrix  $A = (a_{\ell j})_{\ell, j=0}^{n-1} \in \mathbf{R}^{n \times n}$ ,  $n = k^2$ ,  $k \in \mathbf{N}$ , where

$$a_{\ell j} = \begin{cases} 4 & , \ell = j \\ -1 & , (\ell \text{DIV } k) = (j \text{ DIV } k) \wedge |\ell - j| = 1 \\ -1 & , |\ell - j| = k \\ 0 & , \text{otherwise} \end{cases}, \quad 1 \leq \ell, j \leq n-1.$$

Here,  $\ell \text{ DIV } k$  stands for the integer part of  $\ell/k$ , whereas  $\ell \text{ MOD } k \in \{0, \dots, k-1\}$  denotes the rest of the division.

(i) Represent  $A$  as a  $k \times k$  block matrix and specify the individual blocks.

(ii) A vector  $x = (x_0, \dots, x_{n-1})^T \in \mathbf{R}^n$  can be stored in a  $k \times k$  array  $\hat{x}$  as follows:

$$x_\ell = \hat{x}[\ell \text{ DIV } k, \ell \text{ MOD } k], \quad 0 \leq \ell \leq n-1.$$

Using this 'two-dimensional' ordering of the components of the vector, the action of the matrix  $A$  can be easily explained. Describe by means of a two-dimensional drawing how the components of  $\hat{x}$  change under the mapping represented by the matrix  $A$ .

[Hint:  $\hat{x}$  is a  $k \times k$  array of certain values. In case of multiplication by  $A$ , a new value is assigned to each grid point which can be written as a weighted sum of other values. The weights in this sum correspond to the entries of  $A$ .]

(iii) Assume that for  $1 \leq \mu, \nu \leq k$  the vectors  $e_{\mu, \nu} \in \mathbf{R}^n$  are stored in the  $k \times k$  arrays  $\hat{e}_{\mu, \nu}$  of complex numbers. In this form, they can be represented according to

$$\hat{e}_{\mu, \nu}[i, j] := \sin\left(\pi \mu \frac{i+1}{k+1}\right) \sin\left(\pi \nu \frac{j+1}{k+1}\right) \quad , \quad 0 \leq i, j \leq k-1 \quad .$$

Show that  $e_{\mu, \nu}$  is an eigenvector of  $A$  corresponding to the eigenvalue

$$\lambda_{\mu, \nu} = 4\left(\sin^2\left(\frac{\pi}{2} \frac{\mu}{k+1}\right) + \sin^2\left(\frac{\pi}{2} \frac{\nu}{k+1}\right)\right) \quad .$$

(iv) Compute  $\lambda_{max}(A) := \max\{|\lambda|; \lambda \in \sigma(A)\}$ ,  $\lambda_{min}(A) := \min\{|\lambda|; \lambda \in \sigma(A)\}$  and the spectral condition number

$$\kappa(A) := \frac{\lambda_{max}(A)}{\lambda_{min}(A)} \quad .$$

(v) Show that  $A$  is positive definite.

(vi) Derive an estimate for the convergence rate of the cg method in the  $A$  energy norm. How does the convergence behave in terms of  $k$ ?

**Delivery of the homework at latest on October 2, 2009. The homework may be submitted either electronically (rohop@math.uh.edu) or as a hardcopy in class.**