

Exercise Sheet 4

(*) questions contribute, Deadline 3pm 4th December 2006

1. (*) Let the CG iteration be applied to a positive definite matrix problem

$$A\mathbf{x} = \mathbf{b}.$$

While $\mathbf{r}_{k-1} \neq \mathbf{0}$, show that the algorithm proceeds without divisions by zero and we have

$$\mathcal{K}_k := \langle \mathbf{r}_0, \mathbf{r}_1, \dots, \mathbf{r}_{k-1} \rangle = \langle \mathbf{b}, A\mathbf{b}, \dots, A^{k-1}\mathbf{b} \rangle.$$

2. For a positive definite matrix $A \in \mathbb{R}^{n \times n}$ show that $F_A(\mathbf{x}) := \sqrt{\mathbf{x}^T A \mathbf{x}}$ is a vector norm on \mathbb{R}^n .
3. Apply two iterations of the conjugate gradient method for the problem

$$A\mathbf{x} = \mathbf{b}, \text{ where } A = \begin{bmatrix} 2 & -1 & 0 \\ -1 & 2 & -1 \\ 0 & -1 & 2 \end{bmatrix} \text{ and } \mathbf{b} = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix},$$

to obtain \mathbf{x}_2 . Calculate the residual at this iteration and comment on your answer.

4. For a given $A \in \mathbb{R}^{n \times n}$ and $\mu \in \mathbb{R}$, we may write the inverse iteration in the form: Given $\mathbf{x}^0 \in \mathbb{C}^n$ for $k = 1, 2, \dots$
Solve

$$\begin{aligned} (A - \mu I)\mathbf{w} &= \mathbf{x}^{k-1} \\ \mathbf{x}^k &= \mathbf{w} / \|\mathbf{w}\| \\ \lambda^k &= (\mathbf{x}^k)^* A \mathbf{x}^k \end{aligned}$$

Assume A is complete and $\mathbf{x}^0 = \sum_{i=1}^n \alpha_i \mathbf{z}_i$, where $\{\mathbf{z}_i\}$ are the eigenvectors of A with corresponding eigenvalues $\{\lambda_i\}$. Under certain conditions, that are to be stated,

- (a) Describe the eigenvalues and eigenvectors of $(A - \mu I)^{-1}$ in terms of μ and the eigenvalues, eigenvectors of A .
- (b) Show that

$$\mathbf{x}^r = \frac{1}{K_0 K_1 K_2 \cdots K_{r-1}} \sum_{i=1}^n \alpha_i \left(\frac{1}{\lambda_i - \mu} \right) \mathbf{z}_i,$$

where $K_j = \|(A - \mu I)^{-1} \mathbf{x}^j\|$.

- (c) For any $1 \leq J \leq n$, show that

$$\|K_0 K_1 K_2 \cdots K_{r-1} (\lambda_J - \mu)^r \mathbf{x}^r - \alpha_J \mathbf{z}_J\| \leq \left| \frac{\lambda_J - \mu}{\lambda_I - \mu} \right|^r \sum_{i=1, i \neq J}^n |\alpha_i| \|\mathbf{z}_i\|,$$

where the properties of the index I are to be stated.

- (d) Describe how a suitable choice of μ may produce a fast convergence of \mathbf{x}^r to an element in the subspace spanned by the eigenvector \mathbf{z}_J .

5. Let

$$A = \begin{bmatrix} 3 & 2 & 1 \\ 2 & 4 & 1 \\ 1 & 1 & 2 \end{bmatrix}, \quad \mathbf{b} = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}.$$

- (a) Apply three iterations of the power method to A to find the largest eigenvalue.

- (b) Apply three iterations of the inverse iteration, with $\mu = 1.3$ to find the smallest eigenvalue of A .
- (c) Apply three iterations of the Rayleigh quotient iteration, with a suitable initial μ to find the middle eigenvalue of A .

6. (*) **Computation/MATLAB**

- (a) By setting the random number generator seed to 20 construct a 200 by 200 random matrix, B . Setting the seed to 50 construct a column vector b_{200} of length 200.
- (b) Now construct

$$A_{200} = I + B^t B.$$

Calculate the largest and smallest eigenvalues in magnitude of A_{200} .

- (c) Construct the Conjugate Gradient function given in the lecture notes. Setting $tol = 10^{-6}$ find the number of iterations required for the conjugate gradient method to obtain this tolerance.
- (d) Repeat the previous three steps but with a 400 by 400 matrix, A_{400} , and then an 800 by 800 matrix, A_{800} . Comment on the results.