TKK Matematiikan laitos

Mat-1.3651 Numerical Linear Algebra, spring 2008 (Numeerinen matriisilaskenta, kevät 2008)

Exercise 10 (10.4.2008)

Please hand in the exercises marked with an asterisk (*) either to the assistant's folder in front of U313 or latest at the beginning of the exercise. This is the last exercise. You may take the course exam on Thursday 17th April, 2008 at 2.30pm in lecture hall G. In the April exam 25% of the grade is determined by the hand-in home assignments.

Another possibility to take the exam is on Saturday 10th May, 2008. Check the official exam schedules for the correct location. If you take the May exam, you must specify whether or not your home assignments affect the grade.

- * 1. Suppose that in the Arnoldi iteration (given A and b) an entry $h_{n+1,n} = 0$ is encountered.
 - (a) Show how the formula $AQ_n = Q_{n+1}\tilde{H}_n$ can now be simplified. What does this imply about the structure of a full $m \times m$ Hessenberg reduction $A = QHQ^*$ of A?
 - (b) Show that \mathcal{K}_n is an invariant subspace of A.
 - (c) Show that the Krylov subspaces fulfil $\mathcal{K}_n = \mathcal{K}_{n+1} = \mathcal{K}_{n+2} = \dots$.
 - (d) Show that $\Lambda(H_n) \subset \Lambda(A)$, i.e. each eigenvalue of H_n is an eigenvalue of A.
 - (e) Show that if A is nonsingular, then the solution x to the system Ax = b lies in \mathcal{K}_n .
 - 2. Let $A \in \mathbb{C}^{m \times m}$ and $b \in \mathbb{C}^m$ be given, and K_n the related $m \times n$ Krylov matrix with K_n^+ the pseudoinverse. Let p_n be the characteristic polynomial of the Hessenberg H_n in the Arnoldi iteration. How is p_n seen in the matrix $K_n^+ A K_n$?

* 3. Let
$$A = \begin{pmatrix} 0 & 1 \\ I_{m-1} & 0 \end{pmatrix} \in \mathbb{R}^{m \times m}$$
 and $b = e_1 \in \mathbb{R}^m$. Compute the Ritz values.

4. GMRES has the least squares problem

$$\|\tilde{H}_n y - \|b\|e_1\| = \min$$

to solve.

- (a) Describe an $O(n^2)$ algorithm based on QR factorization by Givens rotations (see Exercise 5).
- (b) Show how the operation count can be improved to O(n), as mentioned in the lecture, if the problem for step n-1 has already been solved.
- * 5. We know that CG is (also) an iterative minimization of the function $\varphi(x) = \frac{1}{2}x^T A x x^T b, x \in \mathbb{R}^m$. Another way to minimize the same function is the method of *steepest descent*.
 - (a) Derive the formula $\nabla \varphi(x) = -r$. The steepest descent iteration corresponds to the choice $p_n = r_n$ instead of $p_n = r_n + \beta_n p_{n-1}$ in CG.
 - (b) Determine the formula for the optimal step length α_n of the steepest descent iteration.
 - (c) Write down the full steepest descent iteration. There are three operations inside the main loop.