## Advanced Numerical Linear Algebra AMSC/CMSC 763

Homework 1 Due September 18, 2019

1. Let  $\langle f, g \rangle$  denote an inner product for functions defined on a real interval [a, b]. Typical examples have the form

$$\langle f,g \rangle \equiv \int_{a}^{b} f(x)g(x)w(x)dx$$

where w(x) is a positive weight function on (a, b). For example, the choice  $w(x) \equiv 1$  gives the  $\ell_2$  inner product and  $w(x) = 1/\sqrt{1-x^2}$  on [-1,1] gives the Chebyshev inner product.

a. Show that polynomials  $\{\psi_j \mid j = 1, 2, ...\}$  orthogonal with respect to such an inner product can be defined via a three-term recurrence

$$\gamma_{j+1}\psi_{j+1}(x) = x\psi_j(x) - \delta_j\psi_j(x) - \gamma_j\psi_{j-1}(x) \tag{1}$$

such that  $\psi_j$  has degree j - 1,  $\|\psi_j\| = 1$  where  $\|f\| \equiv \langle f, f \rangle^{1/2}$ , and  $\psi_0 = 0$ . b. Show that for  $k+1 \leq n$ , the roots of  $\psi_{k+1}$  are the eigenvalues of the tridiagonal matrix  $T_k$  determined by the recurrence (1). *Hint:* Consider the characteristic polynomial of  $T_k$ .

c. The expression (1) resembles the recurrence that defines the Lanczos algorithm for a symmetric matrix A of order n,

$$\gamma_{j+1}v_{j+1} = Av_j - \delta_j v_j - \gamma_j v_{j-1},\tag{2}$$

which produces orthogonal vectors  $\{v_i\}$ . Derive a variant of (2) of the form

$$\gamma_{j+1}w_{j+1} = \Lambda w_j - \delta_j w_j - \gamma_j w_{j-1},$$

where  $\Lambda = diag(\lambda_1, \lambda_2, \dots, \lambda_n)$  is the diagonal matrix of eigenvalues of A. d. Show that

$$\langle p,q \rangle \equiv \sum_{i=1}^{n} p(\lambda_i) q(\lambda_i) [w_1]_i^2$$

is an inner product on the space of polynomials defined on  $\mathbb{R}$ . What does this tell you about the polynomials from (1) defined using this inner product?

2. Let Ax = b be a linear system of equations where A is symmetric and positive-definite. Let  $x_k$  be the kth iterate generated by the conjugate gradient method (CG). Show that if  $x_k \neq x$ , then the vectors generated by CG satisfy

(i)  $\langle r_k, p_j \rangle = \langle r_k, r_j \rangle = 0, \qquad j < k,$ (ii)  $\langle Ap_k, p_j \rangle = 0, \qquad j < k,$ 

(ii) 
$$\langle Ap_k, p_j \rangle = 0,$$

(iii)  $\operatorname{span}\{r_0, r_1, \dots, r_{k-1}\} = \operatorname{span}\{p_0, p_1, \dots, p_{k-1}\}$ =  $\mathcal{K}(A, r_0) \equiv \operatorname{span}\{r_0, Ar_0, \dots, A^{k-1}r_0\}.$ 

Hint: Prove (i) and (ii) simultaneously by induction on k, and use a dimensionality argument for (iii).

3. Let Ax = b be as in Problem 2. Starting from an arbitrary initial iterate  $x_0$ , the steepest descent method generates a sequence of iterates  $x_1, x_2, \ldots$  by the computation

$$x_{k+1} = x_k + \alpha_k r_k \,,$$

where  $r_k$  is the residual  $b - Ax_k$  and  $\alpha_k$  is a scalar chosen so that the norm  $||x - x_{k+1}||_A$  is minimal.

a. Explain the name "steepest descent method."

b. Show that the error  $e_k = x - x_k$  satisfies

$$\|e_k\|_A \le \left(\frac{\kappa - 1}{\kappa + 1}\right)^k \|e_0\|_A$$

where  $\kappa = \Lambda/\lambda$  is the *condition number* of A, that is, the ratio of the largest eigenvalue of A to its smallest eigenvalue.

4. A demo (soon to be) given in class shows the effect damped Jacobi smoothing had on the discrete one-dimensional diffusion equation.

a. Implement this demo yourself. That is, show that a few steps of damped Jacobi smoothing makes the error smooth. You can generate the matrix and right-hand side using the code

Reasonable choices for n are 31 or 63, but feel free to play with anything you like. To make the case, start with a random initial value and then plot the error in one or two figures.

b. Continue this experiment by implementing the *two-grid* algorithm. This will require construction of the coarse-grid matrix  $A_{2h}$  and the prolongation and restriction operators, P and R. You can then take one step of the two-grid algorithm to consist of two smoothing steps, followed by restriction, coarse-grid correction, and prolongation. Show that this algorithm displays "textbook" multigrid behavior, that is, the number of steps needed for the error to be smaller than a given tolerance is independent of the discretization mesh size.