

## APPM 5600 — HOMEWORK # 6

1. (15 points) In class we showed that update to the basis for the residual space is given by

$$\mathbf{p}_{k+1} = \mathbf{r}_{k+1} - \frac{\langle \mathbf{p}_k, \mathbf{r}_{k+1} \rangle_{\mathbf{A}}}{\|\mathbf{p}_k\|_{\mathbf{A}}^2} \mathbf{p}_k. \quad (1)$$

In this exercise, we will remove the need to apply  $\mathbf{A}$  to create this update (less computing  $\mathbf{r}_{k+1}$ ).

- (a) (6 points) Using the fact that  $\mathbf{r}_{k+1} = \mathbf{r}_k - \alpha_k \mathbf{A} \mathbf{p}_k$  and  $\mathbf{r}_{k+1}^T \mathbf{r}_k = 0$ , show that  $\langle \mathbf{p}_k, \mathbf{r}_{k+1} \rangle_{\mathbf{A}} = -\frac{\|\mathbf{r}_{k+1}\|_2^2}{\alpha_k}$ .
  - (b) (6 points) Rewrite  $\|\mathbf{p}_k\|_{\mathbf{A}}^2$  in terms of  $\mathbf{r}_k$  and  $\alpha_k$ .
  - (c) (3 points) Plug these expressions into (1) to get a technique for evaluating the next basis vector for the residual space without any applications of the matrix  $\mathbf{A}$ .
2. (50 points) Consider a sparse  $500 \times 500$  matrix  $\mathbf{A}$  constructed as follows.
- Put a 1 in each diagonal entry.
  - In each off-diagonal entry put a random number from the uniform distribution on  $[-1, 1]$  but make sure to maintain symmetry. Then replace each off-diagonal entry with  $|a_{ij}| > \tau$  by 0, where  $\tau$  is a parameter. For  $\tau = 0.01, 0.05, 0.1$ , and  $0.2$ . (You should have 4 matrices  $\mathbf{A}_j$ . One for each  $\tau$ .)

Take the right hand side to be a random vector  $\mathbf{b}$  and set the tolerance to  $10^{-10}$ .

- (a) (10 points) Write a Steepest Descent and Conjugate Gradient solver.
  - (b) (10 points) Apply Steepest Descent to solve each of the linear systems and plot the residual for each iteration  $\|\mathbf{r}_n\|$  versus the iteration  $n$  on a semilogy scale. Put all of these plots on the same graph. linear system.
  - (c) (10 points) Apply Conjugate Gradient to solve each of the linear systems and plot the residual for each iteration  $\|\mathbf{r}_n\|$  versus the iteration  $n$  on a semilogy scale. Put all of these plots on the same graph.
  - (d) (10 points) What do you observe about the convergence of these methods? If the methods do not converge for any choices of  $\tau$  explain what happening.
  - (e) (10 points) How does the residual compare with the error bounds provided in class? (This is strictly comparing the numbers that you get.)
3. (15 points) Suppose Conjugate Gradient is applied to a symmetric positive definite matrix  $\mathbf{A}$  with the result  $\|\mathbf{e}_0\|_{\mathbf{A}} = 1$ , and  $\|\mathbf{e}_{10}\|_{\mathbf{A}} = 2 \times 2^{-10}$ , where  $\|\mathbf{e}_k\|_{\mathbf{A}} = \|\mathbf{x}_k - \mathbf{x}^*\|_{\mathbf{A}}$  and  $\mathbf{x}^*$  is the true solution. Based solely on this data,
- (a) (7 points) What bound can you give on  $\kappa(\mathbf{A})$ ?
  - (b) (8 points) What bound can you give on  $\|\mathbf{e}_{20}\|_{\mathbf{A}}$ ?

Related note: Theorem 38.2 from Trefethen and Bau gives a slightly more interesting bound.

**Thm 38.2** Let the conjugate gradient iteration be applied to a symmetric positive definite matrix problem  $\mathbf{Ax} = \mathbf{b}$ . If the iteration has not already converged (i.e.  $\mathbf{r}_{n-1} \neq \mathbf{0}$ ), then  $\mathbf{x}_n$  is the unique point in the Krylov space that minimizes  $\|\mathbf{e}_n\|_{\mathbf{A}}$ . The convergence is monotonic,

$$\|\mathbf{e}_n\|_{\mathbf{A}} \leq \|\mathbf{e}_{n-1}\|_{\mathbf{A}}$$

4. (20 points) Consider the task of solving the following system of nonlinear equations.

$$f_1(x, y) = 3x^2 + 4y^2 - 1 = 0 \text{ and } f_2(x, y) = y^3 - 8x^3 - 1 = 0$$

for the solution  $\boldsymbol{\alpha}$  near  $(x, y) = (-0.5, 0.25)$ .

- (a) (10 points) Apply the fixed point iteration with  $\mathbf{g}(\mathbf{x}) = \mathbf{x} - \begin{bmatrix} 0.016 & -0.17 \\ 0.52 & -0.26 \end{bmatrix} \begin{bmatrix} 3x^2 + 4y^2 - 1 \\ y^3 - 8x^3 - 1 \end{bmatrix}$ . You can use  $(-0.5, 0.25)$  as the initial condition. How many steps are needed to get an approximation to 7 digits of accuracy?
- (b) (10 points) Why is this a good choice for  $\mathbf{g}(\mathbf{x})$ ?