2021 Spring MAS 365: Homework 5

posted on Apr 8; due by Apr 15

1. [10+10 points] The nonlinear system

$$x_1^2 - 10x_1 + x_2^2 + 8 = 0$$
, $x_1x_2^2 + x_1 - 10x_2 + 8 = 0$

can be transformed into the fixed-point problem

$$x_1 = g_1(x_1, x_2) = \frac{x_1^2 + x_2^2 + 8}{10}, \quad x_2 = g_2(x_1, x_2) = \frac{x_1 x_2^2 + x_1 + 8}{10}.$$

(a) Use Theorem 10.6 in the textbook to show that $G = (g_1, g_2)^t$ mapping $D \subset \mathbb{R}^2$ to \mathbb{R}^2 has a unique fixed point in

$$D = \{(x_1, x_2)^t \mid 0 \le x_1, x_2 \le 1.5\}.$$

(b) Use Theorem 10.6 to estimate the number of iterations required for the fixed-point iteration to achieve 10^{-3} accuracy of $||\boldsymbol{x}^{(k)} - \boldsymbol{p}||_{\infty}$ with $\boldsymbol{x}^{(0)} = (0,0)^t$, where \boldsymbol{p} is the unique fixed point.

Solution:

(a) G is a polynomial, so it is continuous. Since

$$0.8 \le \frac{x_1^2 + x_2^2 + 8}{10} \le 1.25, \quad 0.8 \le \frac{x_1 x_2^2 + x_1 + 8}{10} \le 1.2875$$

for all $x \in D$, we have that $G(x) \in D$ whenever $x \in D$. In addition, since

$$\left| \frac{\partial g_1(\boldsymbol{x})}{\partial x_1} \right| = \left| \frac{2x_1}{10} \right| \le \frac{3}{10}, \quad \left| \frac{\partial g_1(\boldsymbol{x})}{\partial x_2} \right| = \left| \frac{2x_2}{10} \right| \le \frac{3}{10},$$
$$\left| \frac{\partial g_2(\boldsymbol{x})}{\partial x_1} \right| = \left| \frac{x_2^2 + 1}{10} \right| \le \frac{3.25}{10}, \quad \left| \frac{\partial g_1(\boldsymbol{x})}{\partial x_2} \right| = \left| \frac{2x_1x_2}{10} \right| \le \frac{4.5}{10},$$

we have that

$$\left| \frac{\partial g_i(\boldsymbol{x})}{\partial x_j} \right| \le 0.45 = \frac{0.9}{2}$$

for all i, j = 1, 2. Therefore, by Theorem 10.6, G has a unique fixed point in D.

(b) We find an integer k that satisfies

$$||\boldsymbol{x}^{(k)} - \boldsymbol{p}||_{\infty} \le \frac{K^k}{1 - K} ||\boldsymbol{x}^{(1)} - \boldsymbol{x}^{(0)}||_{\infty} < 10^{-3},$$

where K = 0.9 and $\mathbf{x}^{(1)} = (0.8, 0.8)^t$. This reduces to

$$\frac{0.9^k}{0.1}0.8 < 10^{-3}$$

which is equivalent to

$$k \log_{10} 0.9 < -3 - \log_{10} 8$$
 and $k > \frac{-3 - \log_{10} 8}{\log_{10} 0.9} \approx 85.2995$

1

Hence, 86 iterations will ensure the desired accuracy.

2. [10 points] What does Newton's method reduce to for the linear system Ax = b given by

$$a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = b_1,$$

$$a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = b_2,$$

$$\vdots$$

$$a_{n1}x_1 + a_{n2}x_2 + \dots + a_{nn}x_n = b_n,$$

where A is a nonsingular matrix?

Solution: Let $f_j(x_1, \ldots, x_n) = a_{j1}x_1 + \cdots + a_{jn}x_n - b_j$. Since $\frac{\partial f_j}{\partial x_i} = a_{ji}$, the Jacobian matrix is $J(\boldsymbol{x}) = A$. Then, one iteration of Newton's method is as follows:

$$\boldsymbol{x}^{(1)} = \boldsymbol{x}^{(0)} - J(\boldsymbol{x}^{(0)})^{-1}F(\boldsymbol{x}^{(0)}) = \boldsymbol{x}^{(0)} - A^{-1}(A\boldsymbol{x}^{(0)} - \boldsymbol{b}) = A^{-1}.$$

Therefore, for any given $x^{(0)}$, one iteration of Newton's method finds the solution.

- 3. [10+10+10 points]
 - (a) Implement the conjugate gradient method via MATLAB grader.
 - (b) Implement the preconditioned conjugate gradient method via MATLAB grader.
 - (c) Implement Newton's method via MATLAB grader.

Solution:

```
(a) function [xc Nc] = conjugate_gradient(A, b, x0, epsilon, N)
       x = x0;
       r = b - A*x;
       v = r;
       rr0 = r'*r;
       for k=1:N
           xprev = x;
           Av = A*v;
           t = rr0/(v'*Av);
           x = x + t*v;
           r = r - t*Av;
           rr1 = r'*r;
           s = rr1/rr0;
           v = r + s*v;
           rr0 = rr1;
           if (norm(x - xprev, 2)/norm(x, 2) < epsilon)
                break;
            end
       end
       xc = x;
       Nc = k;
   end
(b) function [xp Np] = preconditioned_conjugate_gradient(A, b, x0, epsilon, N)
       D = diag(diag(A));
       sqrt_invD = sqrt(inv(D));
       Ap = sqrt_invD*A*sqrt_invD;
       bp = sqrt_invD*b;
```

```
x = sqrt(D)*x0;
       r = bp - Ap*x;
       v = r;
       rr0 = r'*r;
       for k=1:N
           xprev = x;
           Av = Ap*v;
           t = rr0/(v'*Av);
           x = x + t*v;
           r = r - t*Av;
           rr1 = r'*r;
           s = rr1/rr0;
           v = r + s*v;
           rr0 = rr1;
            if (norm(sqrt_invD*(x - xprev), 2)/norm(sqrt_invD*x,2) < epsilon)</pre>
                break;
           end
       end
       xp = sqrt_invD*x;
       Np = k;
(c) function sol = newton(x0, N, epsilon)
       x = x0;
       for k=1:N
           F = [3*x(1) - \cos(x(2)*x(3)) - 1/2;
                 4*x(1)^2 - 625*x(2)^2 + 2*x(2) - 1;
                 \exp(-x(1)*x(2)) + 20*x(3) + (10*pi-3)/3];
            J = [3 x(3)*sin(x(2)*x(3)) x(2)*sin(x(2)*x(3));
                 8*x(1) -1250*x(2)+2 0;
                 -x(2)*exp(-x(1)*x(2)) -x(1)*exp(-x(1)*x(2)) 20];
           x_prev = x;
           x = x - J \setminus F;
            if (norm(x - x_prev, Inf)/norm(x, Inf) < epsilon)</pre>
                break;
            end
       end
       sol = [x; k];
   end
```