Problem Set 3 Numerical Analysis II

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Problem 1. Following the notions from the lecture notes, carefully prove the convergence of the following projected gradient method with a variable step-length:

$$u_{n+1} = P_K[u_n - \alpha_n \nabla F(u_n)].$$

Give an example of $\{\alpha_n\}$ satisfying the hypothesis that you need for the convergence.

Solution. Let $K \subset \mathbb{R}^m$ be nonempty, closed, and convex. Let $\nabla F: \mathbb{R}^m \to \mathbb{R}^m$ be L-Lipschitz continuous and m-strongly monotone, and let \bar{u} be the unique solution to the stationary problem. Moreover, let $\bar{\alpha} \in (0, (2m)/L^2)$ satisfy

$$\bar{u} = P_K \left[\bar{u} - \bar{\alpha} \nabla F(\bar{u}) \right].$$

Supposing that $\lim_{n\to\infty} \alpha_n \to \bar{\alpha}$, since P_K is a nonexpansive mapping, we have

$$||u_{n+1} - \bar{u}||^2 \le ||u_n - \bar{u}||^2 - 2\langle \alpha_n \nabla F(u_n) - \bar{\alpha} \nabla F(\bar{u}), u_n - \bar{u} \rangle + ||\alpha_n \nabla F(u_n) - \bar{\alpha} \nabla F(\bar{u})||^2$$

$$\approx ||u_n - \bar{u}||^2 - 2\bar{\alpha} \langle \nabla F(u_n) - \nabla F(\bar{u}), u_n - \bar{u} \rangle + \bar{\alpha}^2 ||\nabla F(u_n) - \nabla F(\bar{u})||^2.$$

By the assumptions above we have

$$\|\nabla F(u_n) - \nabla F(\bar{u})\| \le L\|u_n - \bar{u}\|,$$
$$\langle \nabla F(u_n) - \nabla F(\bar{u}), u_n - \bar{u} \rangle \ge m\|u_n - \bar{u}\|,$$

which imply that

$$||u_{n+1} - \bar{u}||^2 \le (1 - 2\alpha m + \alpha^2 L^2) ||u_n - \bar{u}||^2.$$

Letting $\Theta = \sqrt{1 - 2\alpha m + \alpha^2 L^2}$, we end up with

$$||u_{n+1} - \bar{u}|| \le \Theta ||u_n - \bar{u}||,$$

from which we can conclude the convergence $\{u_n\} \to \bar{u}$ from the Banach Contraction Principle.

As an example of a sequence $\{\alpha_n\}$ satisfying the hypothesis, consider

$$\alpha_N = \frac{2m}{L^2} - \varepsilon_N,\tag{1}$$

where $N \in \mathbb{Z}$ and ε_N satisfies $|\varepsilon_N| < \varepsilon$ for some arbitrarily small $\varepsilon > 0$. Then, from (1) we have

$$\left|\alpha_N - \frac{2m}{L^2}\right| = |\varepsilon_N| < \varepsilon,$$

so our choice of $\{\alpha_n\}$ satisfies the hypothesis, as desired.

Problem 2. Let n be an even integer, and consider the $n \times n$ matrix A with 3 on the main diagonal, -1 on the super- and sub-diagonal, and 1/2 in the (i, n + 1 - i) position for all i = 1, ..., n except for i = n/2 and n/2 + 1. Define a vector $\mathbf{b} = [2.5, 1.5, ..., 1.5, 1.0, 1.0, 1.5, ..., 1.5, 2.5]^T$ where there are n - 4 repetitions of 1.5 and two repetitions of 1. Write MATLAB codes and solve the system $A\mathbf{x} = \mathbf{b}$ for n = 100 (a system of 100 equations with 100 unknowns) using the following methods:

- a) Steepest Descent with a fixed step length.
- b) Steepest Descent with a variable step length.
- c) Conjugate Gradient.

Solution. The three methods are written in the following snippet:

```
% Steepest Descent with fixed step length
  function [x,mag,iter] = sdf(A,b,x0,tol,it_max,alpha)
      r0 = b - A*x0;
      x = x0;
      for i = 1:it_max
          x = x + alpha*r0;
          r0 = b - A^*x;
          mag = norm(r0);
           iter = i;
           if mag < tol</pre>
10
               break
11
           end
12
      end
13
14 end
17 % Steepest Descent with variable step length
function [x,mag,iter] = sdv(A,b,x0,tol,it_max)
19
      r0 = b - A*x0;
      x = x0;
20
21
      for i = 1:it_max
          alpha = (r0'*r0)/(r0'*A*r0);
22
          x = x + alpha*r0;
23
24
          r0 = b - A^*x;
          mag = norm(r0);
25
           iter = i;
26
           if mag < tol
27
               break
28
           end
29
      end
30
31 end
34 % Conjugate Gradient method
  function [x,mag,iter] = cgm(A,b,x0,tol,it_max)
      r0 = b - A^*x0;
36
      d = r0;
37
      x = x0;
38
      for i = 1:it_max
39
40
          w = A^*d;
          alpha = (r0'*r0)/(d'*w);
41
          x = x + alpha*d;
42
          mag2 = norm(r0);
43
          r0 = r0 - alpha*w;
44
          mag = norm(r0);
45
           iter = i;
46
           if mag < tol</pre>
47
48
               break
49
          beta = (mag^2)/(mag^2);
51
           d = r0 + beta*d;
52
53 end
```

We now implement these methods for the given *A* and **b**:

```
% Form matrix A
  function A = makeA(n)
      if mod(n,2) \sim 0
          A = eye(1);
      else
          A = 3*_{eye}(n);
           for i = 1:n-1
               A(i+1,i) = -1;

A(i,i+1) = -1;
               A(i,n+1-i) = 1/2;
11
          A(n,1) = 1/2;
          A(n/2,n+1-(n/2)) = -1;
13
          A((n/2)+1,n-(n/2)) = -1;
14
      end
15
16 end
  % Form vector b
18
  function b = makeb(n)
19
      b = ones(n,1);
20
21
      b(1) = 2.5;
      b(2:(n/2)-1) = 1.5;
22
      b((n/2)+2:n-1) = 1.5;
23
24
      b(n) = 2.5;
25
27
  function [A,b] = makeAb(n)
      A = makeA(n);
30
      b = makeb(n);
  end
31
33 % Variables
_{34} [A,b] = makeAb(100);
35 tol = 1e-4;
it_max = 1000;
x = 5 + zeros(100, 1);
39 % Steepest descent with fixed step length
  [x1,n1,iter1] = sdf(A,b,x,tol,it_max,0.001);
% Steepest descent with variable step length
  [x2,n2,iter2] = sdv(A,b,x,tol,it_max);
45 % Conjugate Gradient method
[x3,n3,iter3] = cgm(A,b,x,tol,it_max);
```

Problem 3. Find the DFT of the vector $\mathbf{x} = [0, 1, 0, -1, 0, 1, 0, -1]^{T}$.

Proof. Let $y = [y_0, ..., y_{n-1}]^T$, where the k^{th} element is given by the expression

$$y_k = \frac{1}{\sqrt{n}} \sum_{j=0}^{n-1} x_j e^{-\frac{i2\pi jk}{n}}.$$

Here we use ι to denote the imaginary number $\sqrt{-1}$. Now, since in this case n = 8, we have

$$\omega^{j} := e^{-(2\pi \iota j)/8} = \cos \frac{\pi}{4} j - \iota \sin \frac{\pi}{4} j.$$

Thus, we collect the ω^i :

$$\omega^0=1,\quad \omega^1=\frac{1-\iota}{\sqrt{2}},\quad \omega^2=\iota,\quad \omega^3=\frac{-1-\iota}{\sqrt{2}},\quad \omega^4=-1,\quad \omega^5=\frac{\iota-1}{\sqrt{2}},\quad \omega^6=\iota,\quad \omega^7=\frac{1+\iota}{\sqrt{2}}.$$

That leads us to the system

$$\mathbf{y} = \frac{1}{2\sqrt{2}} \begin{bmatrix} \omega^{0} & \omega^{0} \\ \omega^{0} & \omega^{1} & \omega^{2} & \omega^{3} & \omega^{4} & \omega^{5} & \omega^{6} & \omega^{7} \\ \omega^{0} & \omega^{2} & \omega^{4} & \omega^{6} & \omega^{0} & \omega^{2} & \omega^{4} & \omega^{6} \\ \omega^{0} & \omega^{3} & \omega^{6} & \omega^{1} & \omega^{4} & \omega^{7} & \omega^{2} & \omega^{5} \\ \omega^{0} & \omega^{4} & \omega^{0} & \omega^{4} & \omega^{0} & \omega^{4} & \omega^{0} & \omega^{4} \\ \omega^{0} & \omega^{5} & \omega^{2} & \omega^{7} & \omega^{4} & \omega^{1} & \omega^{6} & \omega^{3} \\ \omega^{0} & \omega^{6} & \omega^{4} & \omega^{2} & \omega^{0} & \omega^{6} & \omega^{4} & \omega^{2} \\ \omega^{0} & \omega^{7} & \omega^{6} & \omega^{5} & \omega^{4} & \omega^{3} & \omega^{2} & \omega^{1} \end{bmatrix} \begin{bmatrix} 0 \\ 1 \\ 0 \\ -1 \\ 0 \\ 1 \\ 0 \\ -1 \end{bmatrix}$$

$$= \begin{bmatrix} 0 \\ 0 \\ -\sqrt{2}\iota \\ 0 \\ 0 \end{bmatrix}.$$

Problem 4. Find the inverse DFT of the vector $\mathbf{x} = [1, -i, 1, i]^{\mathsf{T}}$.

Proof. We want to find $\mathbf{x} = F_n^{-1}(\mathbf{y})$, where F_n^{-1} is the inverse DFT. This time the ω^j are given by

$$\omega^j := e^{-(2\pi \iota j)/4} = \cos \frac{\pi}{2} j - \iota \sin \frac{\pi}{2} j.$$

Hence we end up with

$$\mathbf{x} = \frac{1}{2} \begin{bmatrix} (\omega^{0})^{-1} & (\omega^{0})^{-1} & (\omega^{0})^{-1} & (\omega^{0})^{-1} \\ (\omega^{0})^{-1} & (\omega^{1})^{-1} & (\omega^{2})^{-1} & (\omega^{3})^{-1} \\ (\omega^{0})^{-1} & (\omega^{2})^{-1} & (\omega^{0})^{-1} & (\omega^{2})^{-1} \\ (\omega^{0})^{-1} & (\omega^{3})^{-1} & (\omega^{2})^{-1} & (\omega^{1})^{-1} \end{bmatrix} \begin{bmatrix} 1 \\ -\iota \\ 1 \\ \iota \end{bmatrix}$$

$$= \frac{1}{2} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & \iota & -1 & -\iota \\ 1 & -1 & 1 & -1 \\ 1 & -\iota & -1 & \iota \end{bmatrix} \begin{bmatrix} 1 \\ -\iota \\ 1 \\ \iota \end{bmatrix}$$

$$= \begin{bmatrix} 1 \\ -\iota \\ 1 \end{bmatrix}.$$

Problem 5. Write down all primitive seventh roots of unity.

Proof. Since the GCD (greatest common divisor) between 7 and k, where $1 \le k < 7$, is 1 (i.e., 7 is prime), the primitive roots of unity are

$$e^{\frac{2\pi}{7}\iota}, e^{\frac{4\pi}{7}\iota}, e^{\frac{6\pi}{7}\iota}, e^{\frac{8\pi}{7}\iota}, e^{\frac{8\pi}{7}\iota}, e^{\frac{10\pi}{7}\iota}, e^{\frac{12\pi}{7}\iota}.$$