

E0 230 : Computational Methods of Optimization  
Assignment 1 (Due: 29<sup>th</sup> November)

**Instructions:**

- Attempt all questions
- For Numerical Answer questions, insert only the digits **with no leading or trailing spaces** in the MS Teams Form provided
- You can submit the MS Teams Form **only once**. No further changes can be made, so click on *Submit* wisely.
- Late Submissions will be penalised.

1. Consider a matrix  $B \in \mathbb{R}^{m \times n}$ . Assume  $\sigma_1 > \sigma_2 > \dots > \sigma_n$ . Consider the following claims. Select **all** of the statements that hold.

A.  $\sigma_1 \leq \|B\|_F$

B.  $\sigma_{k+1} \leq \frac{\|B\|_F}{\sqrt{k+1}}$

C.  $\sigma_{k+1} \leq \frac{\|B\|_F}{k+1}$

**Solution:** Obviously,  $\sigma_1 \leq \sqrt{\sum_i \sigma_i^2}$ . So A is true. From Tutorial 1, we saw that

$$\arg \min_{\hat{B} \text{ of rank } k+1} \|B - \hat{B}\|_F = B_{k+1} = \sum_{i \leq k+1} \sigma_i v_i u_i^T$$

Now, we know that

$$\begin{aligned} \|B_{k+1}\| &= \sqrt{\sigma_1^2 + \dots + \sigma_{k+1}^2} \\ &\geq \sqrt{\sigma_{k+1}^2 (k+1)} \\ &\geq \sqrt{k+1} \sigma_{k+1} \end{aligned}$$

But  $\|B_{k+1}\|_F \leq \|B\|_F \Rightarrow \|B\|_F \geq \sqrt{k+1} \sigma_{k+1}$ . Thus B is true.

2. Suppose a quadratic function  $(x^T Q x)$  is expanded as:

$$f(x_1, x_2, x_3) = x_1^2 + x_2^2 + 5x_3^2 + 2\varepsilon x_1 x_2 - 2x_1 x_3 + 4x_2 x_3$$

Determine the range of values  $\varepsilon$  can take so that  $Q \succ 0$ .

A.  $\varepsilon \in (-\infty, 0) \cap (\frac{4}{5}, +\infty)$  (i.e. No  $Q \succ 0$  is possible for any  $\varepsilon$ )

B.  $\varepsilon \in (-1, 1)$

C.  $\varepsilon \in (-\frac{4}{5}, 0)$

D.  $\varepsilon \in (-\infty, +\infty)$  (i.e.  $Q \succ 0$  for all values of  $\varepsilon$ )

**Solution:** We represent  $f(x)$  in quadratic form  $x^T Q x$  where

$$Q = \begin{bmatrix} 1 & \varepsilon & -1 \\ \varepsilon & 1 & 2 \\ -1 & 2 & 5 \end{bmatrix}$$

To have  $Q \succ 0$ , all the leading principal minors of Q must be positive (*Sylvester's criterion*). The leading principal minors of Q are  $\Delta_1 = 1, \Delta_2 = 1 - \varepsilon^2$ , and  $\Delta_3 = -5\varepsilon^2 - 4\varepsilon$ . Solving inequalities  $\Delta_1 > 0, \Delta_2 > 0$ , and  $\Delta_3 > 0$ , we get  $\varepsilon \in (-\frac{4}{5}, 0)$ .

3. Let  $f(x) : \mathbb{R}^n \rightarrow \mathbb{R}$  be continuous. It is known that if  $f(x)$  is coercive and first partial derivatives of  $f(x)$  exist on all of  $\mathbb{R}^n$ , then global minimizers of  $f(x)$  exist and is within the critical points of  $f(x)$ .

Given  $h(x, y) = x^4 - 4xy + y^4$ . We wish to find the global minimum by locating the local minima (by computing the gradient and the hessian of  $h$ ). Then,

- (a)  $h$  has global minimum within the local minima because:

- A. Hessian is positive everywhere.
  - B.  $h$  is coercive.**
  - C.  $h$  is bounded from below.
  - D. The global minimum does not exist within the local minima.
- (b) Choose **all** of the statements that are true for  $h$  from the following choices:
- A.  $(0, 0)$  is the only critical point and a local minimum.
  - B.  $(0, 0), (1, 1), (-1, -1)$  are the complete set of real critical points.**
  - C.  $(0, 0), (-1, -1)$  are local minimum.
  - D. Hessian at  $(1, 1), (-1, -1)$  are positive.**
- (c) The global minimum of  $h$  is **-2**

**Solution:** It is easy to see by taking derivative and checking the Hessian that the local minimum of the function is indeed at -2. We now prove  $h$  is coercive to justify that -2 is also the global minimum.

Note that  $h(x, y)$  can be written as,

$$h(x, y) = x^4 + y^4 \left( 1 - \frac{4xy}{x^4 + y^4} \right)$$

As  $\|(x, y)\| = \sqrt{x^2 + y^2} \rightarrow \infty$ , the term  $\frac{4xy}{x^4 + y^4} \rightarrow 0$ . Hence,

$$\lim_{\|(x, y)\| \rightarrow \infty} h(x, y) = \lim_{\|(x, y)\| \rightarrow \infty} (x^4 + y^4) = +\infty$$

. Thus,  $h$  is coercive.

4. Consider the following optimization problem.

$$x^* = \arg \min \frac{1}{2} x^T Q x - b^T x \quad (1)$$

where

$$Q = \begin{bmatrix} 2.3346 & 1.1384 & 2.5606 & 1.4507 \\ 1.1384 & 0.7860 & 1.2743 & 0.9531 \\ 2.5606 & 1.2743 & 2.8147 & 1.6487 \\ 1.4507 & 0.9531 & 1.6487 & 1.8123 \end{bmatrix} \text{ and } b = \begin{bmatrix} 0.4218 \\ 0.9157 \\ 0.7922 \\ 0.9595 \end{bmatrix}$$

We aim to solve this problem using gradient descent, initialized with

$$x_0 = \begin{bmatrix} -135.1150 \\ -4.5224 \\ 130.1168 \\ -5.6879 \end{bmatrix}.$$

- (a) How many iterations are needed to reach a point  $z$  such that  $\|z - x^*\| \leq 0.0001$  with a stepsize  $\alpha = 0.2$ ? **10 ± 1**
- (b) How many iterations are needed to reach a point  $z$  such that  $\|z - x^*\| \leq 0.0001$  with a stepsize  $\alpha = 0.07$ ? **14 ± 1**
- (c) Repeat the above for stepsizes  $\alpha_1 = 0.0728$  and  $\alpha_2 = 0.2185$ . How many iterations are needed to reach a point  $z$  such that  $\|z - x^*\| \leq 0.0001$  with both the stepsizes? (Enter only one number) **14 ± 1**

- (d) What is the  $\alpha^*$  (correct to 4 significant digits) such that the exact solution is reached in one step with the given choice of  $x_0$ ; that is,  $x^* - x_0 = -\alpha^*(Qx_0 - b)$ ? **0.1457  $\pm$  0.0001**

**Solution:**  $\alpha^* = .1457$ .

- (e) Is there a minimum stepsize for which convergence to the minimum is not guaranteed? If yes, what are the  $10^{100000000}$ th and  $10^{100000000} + 1$ th iterates for that stepsize?

- A.  $\alpha = 2/\lambda_1, x^* + v_1, x^* - v_1$   
 B.  $\alpha = 2/\lambda_1, x^* - v_1, x^* + v_1$   
 C. No such stepsize

**Solution:** First, recall that  $x_0 = x^* + v_1$ . Plugging this into the gradient descent update, we get  $x_1 = (1 - \alpha\lambda)v_1 + x^*$ , and, by induction, we get  $x_k = x^* + (1 - \alpha\lambda)^k v_1$ . Then, recall that

$$\begin{aligned} E_k &= (x_k - x^*)^T Q (x_k - x^*) \\ &= (1 - \alpha\lambda)^{2k}. \end{aligned}$$

This quantity converges if  $\alpha \in (0, \frac{2}{\lambda})$ . At  $\alpha = \frac{2}{\lambda}$   $E_k = 1$  for all  $k$ . Substituting this value of  $\alpha$  back into the update equation that we derived earlier, we get  $x_{k+1} = x^* + (-1)^k v_1$ , after which, see the solution for even and odd terms is  $x^* + v_1$  and  $x^* - v_1$  respectively.

5. We've provided a file which takes a 2d vector as an input, and returns the function value, the gradient, and the Hessian of a 2d function. Consider the initial point  $[x_0, y_0] = [1.2, 1.2]^T$ . We aim to use iterative methods to solve  $\arg \min f(x, y)$ . In the following questions, report the minimum number of iterations required to reach a point  $z$ , where  $\|\nabla f(z)\| \leq \epsilon$ ? (Write  $-1$  to indicate "Does not converge with this stepsize")

- (a) Gradient descent with stepsize of 0.001, and  $\epsilon = 0.1$  **5  $\pm$  1**  
 (b) Gradient descent with stepsize of 0.001, and  $\epsilon = 0.01$  **6146  $\pm$  1**  
 (c) Gradient descent with stepsize of 0.001, and  $\epsilon = 0.001$  **11972  $\pm$  1**  
 (d) Gradient descent with stepsize of 0.001, and  $\epsilon = 0.0001$  **17742  $\pm$  1**  
 (e) Gradient descent with stepsize of 0.002, and  $\epsilon = 0.1$  **70  $\pm$  1**  
 (f) Gradient descent with stepsize of 0.002, and  $\epsilon = 0.01$  **1518  $\pm$  1**  
 (g) Gradient descent with stepsize of 0.002, and  $\epsilon = 0.001$  **4370  $\pm$  1**  
 (h) Gradient descent with stepsize of 0.002, and  $\epsilon = 0.0001$  **7248  $\pm$  1**  
 (i) Gradient descent with stepsize of 0.005, and  $\epsilon = 0.1$  **-1**  
 (j) Gradient descent with stepsize of 0.005, and  $\epsilon = 0.01$  **-1**  
 (k) Gradient descent with stepsize of 0.005, and  $\epsilon = 0.001$  **-1**  
 (l) Gradient descent with stepsize of 0.005, and  $\epsilon = 0.0001$  **-1**  
 (m) Newton's Method with  $\epsilon = 0.1$  **3  $\pm$  1**  
 (n) Newton's Method with  $\epsilon = 0.01$  **3  $\pm$  1**  
 (o) Newton's Method with  $\epsilon = 0.001$  **4  $\pm$  1**  
 (p) Newton's Method with  $\epsilon = 0.0001$  **5  $\pm$  1**

6. For finding the minimum of a function  $f : \mathbb{R} \rightarrow \mathbb{R}$ , four iterative algorithms calculate solutions in the following way ( $x_k$  is the solution in the  $k$ -th iteration):

1.  $x_k = 1 + \frac{1}{k}$
2.  $x_k = 1 + (0.5)^{2^k}$
3.  $x_k = 1 + \frac{1}{k!}$
4.  $x_k = x_{k-1} - 0.5^k, k \geq 1$  (initialised with  $x_0 = 2$ )

- (a) What is the nearest integer to which these algorithms converge? (*Hint: Find  $\lfloor x_k \rfloor$  after large number of iterations*) 1
- (b) Suppose each of the algorithms is executed 10 times. Plot the sequence of solutions obtained on a single graph. What is the correct order in which these algorithms converge (slowest to fastest)?
  - A.  $1 < 2 < 3 < 4$
  - B.  $4 < 1 < 2 < 3$
  - C.  $4 < 3 < 2 < 1$
  - D.  $1 < 4 < 3 < 2$

Solving part (a) and (b) shows that the four algorithms converge to a point, and each of the algorithms converge at a different speed. In general, to understand the **rate of convergence**, we use the following to represent how quickly the error  $e_k = x_k - x^*$  converges to zero:

$$\lim_{n \rightarrow \infty} \frac{|e_{k+1}|}{|e_k|^p} = \lim_{n \rightarrow \infty} \frac{|x_{k+1} - x^*|}{|x_k - x^*|^p} = \mu$$

Here  $p \geq 1$  is called the order of convergence, the constant  $\mu$  is the rate of convergence. This expression may be better understood when it is interpreted as  $|e_{k+1}| = \mu|e_k|^p$  when  $n \rightarrow \infty$ . Obviously, the larger  $p$  and the smaller  $\mu$ , the more quickly the sequence converges. Specially, consider the following cases:

- If  $p = 1$  and
  - if  $\mu = 1$ , the convergence is **sublinear**
  - if  $0 < \mu < 1$ , the convergence is **linear** with the rate of convergence of  $\mu$ .
  - if  $\mu = 0$ , the convergence is **superlinear**
- If  $p = 2$ ,  $|e_{k+1}| = \mu|e_k|^2$ , ( $\mu > 0$ ), the convergence is **quadratic**.

- (c) What is the rate of convergence of Algorithms 1,2,3,4.
  - A. Linear, Sublinear, Superlinear, Quadratic
  - B. Sublinear, Quadratic, Superlinear, Linear
  - C. Superlinear, Quadratic, Sublinear, Linear
  - D. Sublinear, Linear, Superlinear, Quadratic

**Solution:** First, express  $x_k$  in Algorithm 4 only in terms of  $k$ . Note that,

$$\begin{aligned}
 x_{k+1} &= x_k - 0.5^{k+1} \\
 &= x_{k-1} - 0.5^k - 0.5^{k+1} \\
 &\vdots \\
 &= x_0 - (0.5 + \cdots + 0.5^k + 0.5^{k+1}) \\
 &= x_0 - 1 + 0.5^{k+1} && \text{(sum of G.P. Series)} \\
 &= 1 + 0.5^{k+1}
 \end{aligned}$$

Clearly,  $\lim_{k \rightarrow \infty} x_k$  for all 4 sequences is 1. Therefore, the minimizer is,  $x^* = 1$ .

**1:**  $\lim_{k \rightarrow \infty} \frac{|x_{k+1} - x^*|}{|x_k - x^*|} = \lim_{k \rightarrow \infty} \frac{|x_{k+1} - 1|}{|x_k - 1|} = \lim_{k \rightarrow \infty} \frac{\frac{1}{k+1}}{\frac{1}{k}} = \lim_{k \rightarrow \infty} \frac{1}{1 + \frac{1}{k}} = 1$  (**Sublinear**)

**2:**  $\lim_{k \rightarrow \infty} \frac{|x_{k+1} - 1|}{|x_k - 1|^2} = \lim_{k \rightarrow \infty} \frac{0.5^{2^{k+1}}}{0.5^{2^k \cdot 2}} = 1$  (**Quadratic**)

**3:**  $\lim_{k \rightarrow \infty} \frac{|x_{k+1} - 1|}{|x_k - 1|} = \lim_{k \rightarrow \infty} \frac{\frac{1}{(k+1)!}}{\frac{1}{k!}} = \frac{1}{k+1} = 0$  (**Superlinear**)

**4:**  $\lim_{k \rightarrow \infty} \frac{|x_{k+1} - 1|}{|x_k - 1|} = \lim_{k \rightarrow \infty} \frac{0.5^{k+1}}{0.5^k} = 0.5$  (**Linear**)

Assume  $x^*$  is the minimum value of the function, as obtained in part (a) of this problem. In the following questions, report the minimum number of iterations required to reach point  $z$  such that  $z \leq x^* + \epsilon$ . (Use  $z \leq x^* + \epsilon$  as the stopping criteria, and not  $z - x^* \leq \epsilon$  for better numerical stability.)

- (d) Algorithm 1 with  $\epsilon = 0.1$  **10 ± 1**
  - (e) Algorithm 1 with  $\epsilon = 0.01$  **100 ± 1**
  - (f) Algorithm 1 with  $\epsilon = 0.001$  **1000 ± 1**
  - (g) Algorithm 1 with  $\epsilon = 0.00001$  **10000 ± 1**
  - (h) Algorithm 2 with  $\epsilon = 0.1$  **2 ± 1**
  - (i) Algorithm 2 with  $\epsilon = 0.01$  **3 ± 1**
  - (j) Algorithm 2 with  $\epsilon = 0.001$  **4 ± 1**
  - (k) Algorithm 2 with  $\epsilon = 0.00001$  **5 ± 1**
  - (l) Algorithm 3 with  $\epsilon = 0.1$  **4 ± 1**
  - (m) Algorithm 3 with  $\epsilon = 0.01$  **5 ± 1**
  - (n) Algorithm 3 with  $\epsilon = 0.001$  **7 ± 1**
  - (o) Algorithm 3 with  $\epsilon = 0.00001$  **9 ± 1**
  - (p) Algorithm 4 with  $\epsilon = 0.1$  **4 ± 1**
  - (q) Algorithm 4 with  $\epsilon = 0.01$  **7 ± 1**
  - (r) Algorithm 4 with  $\epsilon = 0.001$  **10 ± 1**
  - (s) Algorithm 4 with  $\epsilon = 0.00001$  **17 ± 1**
7. Bisection Method, which is based on Intermediate Value Theorem, is generally used to find roots of a one-dimensional function. In this problem, we will use Bisection Method to find the minimum of a differentiable one-dimensional function by finding the roots of its derivative within an interval  $[a, b]$ . The pseudo-code for Bisection Method is as follows:

- Input: Interval  $[a, b]$ , tolerance  $tol$ , maximum iterations  $N_{max}$
- Initialize:  $x_{left} = a, x_{right} = b$
- Inside loop:
  - $x_{mid} = \frac{x_{left} + x_{right}}{2}$
  - if  $f(x_{mid}) = 0$  or  $\frac{x_{right} - x_{left}}{2} \leq tol$  or  $N_{max}$  is reached then, **Stop** and return  $(x_{mid}, \text{number of iterations})$
  - if  $\text{sign}(f(x_{mid})) \cdot \text{sign}(f(x_{left})) < 0$  then  $x_{right} = x_{mid}$
  - if  $\text{sign}(f(x_{mid})) \cdot \text{sign}(f(x_{left})) > 0$  then  $x_{left} = x_{mid}$
- Output: number of iterations,  $x_{mid}$

- (a) Implement the following function,  $f(x)$ , and its derivative:

$$f(x) = x^4 - 14x^3 + 60x^2 - 70x$$

Plot  $f(x)$  vs  $x \in [0, 7]$ , with distance between two  $x$  values as 0.001. How many stationary points of  $f(x)$  exist in this interval? **3** (Optional: Also plot  $f'(x)$  vs  $x$  for the same interval. Can you identify the stationary points from this plot? Notice how the signs of derivatives are reversed in the neighborhood of stationary points. )

From the plot above, note that a single minimum exists within  $[0, 3]$ . Since  $f : (a, b) \rightarrow \mathbb{R}$  is differentiable and unimodal, we can use Bisection Method to find the minimum of  $f(x)$  by finding the root of  $f'(x)$ . Implement the Bisection Method as described in the pseudocode above to find the root of  $f'(x)$  in the specified intervals and report the minimum number of iterations for the following input combinations :

- (b) Starting Interval:  $[0, 2]$ , Tolerance=0.01 **8  $\pm$  1**
- (c) Starting Interval:  $[0, 2]$ , Tolerance=0.0001 **15  $\pm$  1**
- (d) Starting Interval:  $[0, 3]$ , Tolerance=0.001 **12  $\pm$  1**
- (e) What is the minimizer of  $f(x)$  (correct to three decimal places) if the Bisection Method is executed for 10 times starting with interval  $[0, 2]$ ?
- A. 0.770
- B. 0.775
- C. 0.779**
- D. 0.781
- (f) What is the minimizer of  $f(x)$  (correct to three decimal places) if the Bisection Method is executed for 20 times starting with interval  $[0, 2]$ ?
- A. 0.770
- B. 0.775
- C. 0.779
- D. 0.781**

**Solution:**

We can also calculate number of iterations analytically.

$$\begin{aligned}
 n &= \left\lceil \log_2 \frac{b-a}{\text{tolerance}} \right\rceil && \text{(where [a,b] is the initial range given)} \\
 &= \left\lceil \log_2 \frac{2}{0.01} \right\rceil && \text{(substituting given values from part (b))} \\
 &= 8
 \end{aligned}$$