## CMO

## Sheet 3 — E0230

## Assignment (Due: 2 January 2021)

## Instructions

- Answer all questions
- For numerical questions, answer without any leading or trailing spaces
- You can submit MS Teams Form only once
- If the value of answer is an **integer**, give answer in box provided as an integer alone **without any decimal point**. For example 0, 10
- Otherwise, if the value of answer is a **decimal number**, give answer in x.yy format **rounded to 2 decimal places**. For example 0.50, 1.50
- Follow this instruction unless otherwise specified in the question
- Late submissions will be penalised

- 1. Consider the problem of minimizing the function  $f(x) = 100(x_2 x_1^2)^2 + (1 x_1)^2$ .
  - (a) Determine the point  $\hat{x} = (\hat{x_1}, \hat{x_2})$  corresponding to  $\nabla f(\hat{x}) = 0$ .
    - (i) (1 point)  $\hat{x}_1 =$ \_\_\_\_\_
    - (ii) (1 point)  $\hat{x}_2 =$
  - (b) (3 points) Is  $\hat{x}$  a local minimum or a maximum for f?
  - (c) Apply Newton's method for two iterations starting from  $x^{(0)} = (0,0)$  and give the resultant value of  $x^{(2)}$ .
    - (i) (2.5 points)  $x_1^{(2)} =$ \_\_\_\_\_
    - (ii) (2.5 points)  $x_2^{(2)} =$
  - (d) (3 points) Does Newton's method converge if the initial point is (100, 100)?
  - (e) (2 points) Does the steepest descent method with fixed step size  $\alpha = 0.05$  and initial point (100, 100) converge?
- 2. (3 points) Let  $f: \mathbb{R} \to \mathbb{R}$  be given by  $f(x) = (x-5)^4$ . Let  $x^{(k)}$  be the sequence obtained by applying Newton's method to this function, with  $x^{(0)} = 10$ , and let  $x^*$  be the global minimum of f. Determine the value of

$$\lim_{k \to \infty} \frac{|x^{(k+1)} - x^*|}{|x^{(k)} - x^*|}$$

- 3. Consider the problem of minimizing  $f(x) = (\sqrt[3]{x})^4$ .
  - (a) (1 point) Determine  $x^* = \operatorname{argmin}_x f(x)$ .
  - (b) (2 points) Determine the maximum value of a such that Newton's method converges when started from the interval [-a, a].
- 4. Apply the DFP Quasi-Newton updates for minimizing the function

$$f(x) = x^{T} \begin{pmatrix} 2 & 1 \\ 1 & 1 \end{pmatrix} x + \begin{pmatrix} 1 \\ -1 \end{pmatrix}^{T} x,$$

starting with  $x^{(0)} = (0 \quad 0)^T$  and  $B^{(0)} = I_{2 \times 2}$ .

- (a) What is the global minimum  $x^*$  of f?
  - (i) (1 point)  $x_1^* =$ \_\_\_\_\_
  - (ii) (1 point)  $x_2^* =$ \_\_\_\_\_\_
- (b) (1 point) How many iterations were required to obtain the result?
- (c) Give the value of  $B^{(1)} = \begin{pmatrix} B_{1,1}^{(1)} & B_{1,2}^{(1)} \\ B_{2,1}^{(1)} & B_{2,2}^{(1)} \end{pmatrix}$ :

- (i) (0.5 points)  $B_{1,1}^{(1)} =$
- (ii) (0.5 points)  $B_{1,2}^{(1)} =$
- (iii) (0.5 points)  $B_{2,1}^{(1)} =$
- (iv) (0.5 points)  $B_{2,2}^{(1)} =$ \_\_\_\_\_\_
- 5. Consider the python functions for a function f(x), its gradient  $\nabla f(x)$  and Hessian inverse  $(\nabla^2 f(x))^{-1}$  given in the files f.pkl, grad\_f.pkl and hessian\_inv.pkl, which take as input numpy arrays of length 2 and give their output as numpy arrays. Apply Newton's method starting from initial value  $x^{(0)} = (0,0)$ , where the step size during each iteration is determined using backtracking, with parameters  $\alpha = 0.1$  and  $\beta = 0.7$ . This means that at each step, an initial step size of t = 1 is chosen and updated as  $t \leftarrow \beta t$ , until it satisfies

$$f(x+tu) \le f(x) + \alpha t \nabla f(x)^T u,$$

where u is the update direction. Iterate for k iterations until the update distance  $||x^{(k)} - x^{(k-1)}|| < \epsilon = 0.001$ .

- (a) What is the final value  $x^{(k)}$  obtained?
  - (i) (2.5 points)  $x_1^{(k)} =$ \_\_\_\_\_\_
  - (ii) (2.5 points)  $x_2^{(k)} =$
- (b) (5 points) How many iterations were required to obtain the result?  $k = \underline{\hspace{1cm}}$
- 6. Apply the conjugate gradient algorithm to minimize the function  $f(x) = \frac{1}{2}x^TQx b^Tx$ , with the following values of Q and b:

$$Q = \begin{pmatrix} 2 & 1 & 1 \\ 1 & 2 & 1 \\ 1 & 1 & 2 \end{pmatrix}, b = \begin{pmatrix} 4 \\ 0 \\ 0 \end{pmatrix},$$

with initial value  $x_0 = (0 \quad 0 \quad 0)^T$ .

- (a) Give the final solution  $x^* = (x_1^* \quad x_2^* \quad x_3^*)$ .
  - (i) (1 point)  $x_1^* =$ \_\_\_\_\_\_
  - (ii) (1 point)  $x_2^* =$ \_\_\_\_\_\_
  - (iii) (1 point)  $x_3^* =$ \_\_\_\_\_\_
- (b) (4 points) How many iterations were required for converging to the solution?
- 7. Apply the conjugate gradient algorithm to minimize the function  $f(x) = \frac{1}{2}x^TQx b^Tx$ , with the following values of Q and b:

$$Q = \begin{pmatrix} 1 & 2 & -1 & 1 \\ 2 & 5 & 0 & 2 \\ -1 & 0 & 6 & 0 \\ 1 & 2 & 0 & 3 \end{pmatrix}, b = \begin{pmatrix} 0 \\ 2 \\ -1 \\ 1 \end{pmatrix},$$

with initial value  $x^{(0)} = \begin{pmatrix} 0 & 0 & 0 & 0 \end{pmatrix}^T$ .

- (a) Give the final solution  $x^* = \begin{pmatrix} x_1^* & x_2^* & x_3^* & x_4^* \end{pmatrix}$ .
  - (i) (0.5 points)  $x_1^* =$ \_\_\_\_\_
  - (ii) (0.5 points)  $x_2^* =$ \_\_\_\_\_\_
  - (iii) (0.5 points)  $x_3^* =$
  - (iv) (0.5 points)  $x_4^* =$ \_\_\_\_\_\_
- (b) (5 points) How many iterations were required for converging to the solution?