

<b>Session :</b>	Septembre 2022.
<b>Année d'étude :</b>	M1
<b>Discipline :</b>	<i>Optimization : Theory and Algorithms</i> (Unité d'Enseignements Complémentaire 2).
<b>Titulaire du cours :</b>	M. Lorenzo BASTIANELLO.
<b>Durée de l'épreuve :</b>	1h30.
<b>Document(s) autorisé(s) :</b>	Calculatrice autorisée. Le téléphone portable n'est pas autorisé comme calculette. Documents interdits, ainsi que tout appareil électronique permettant une connexion à distance.

*Exam of Optimization : Theory and Algorithms (4211)*

**Instructions :**

- A bad and messy presentation may result in negative points.
- Vectors are indicated in bold letters.
- *The score of each exercise is given only as a benchmark and it may change.*

**Exercise 1.** (5 points) Suppose that a differentiable function  $f : \mathbb{R} \rightarrow \mathbb{R}$  is unimodal over an interval  $[a, b]$ . Denote  $x^* \in (a, b)$  its unique local minimum.

1. (1 point) Let  $x \in (a, b)$  and suppose that  $f'(x) > 0$ . Find an interval  $I$  strictly contained in  $[a, b]$  such that  $x^* \in I$ .
2. (3 points) Describe the bisection algorithm to find  $x^*$ .
3. (1 point) Show that the bisection algorithm converges linearly with convergence ratio  $\frac{1}{2}$ .

**Exercise 2.** (5 points) Consider the function  $f(x, y) = (x + y^2)^2$ . At the point  $\mathbf{x}_0 = (1, 0)$  we consider the search direction  $\mathbf{p} = (-1, 1)$ .

1. (1 point) Give the definition of directional derivative.
2. (3 points) Give the definition of descent direction and show that  $\mathbf{p}$  is a descent direction.
3. (1 point) Find all  $\mathbf{x}^*$  that minimize  $f$ .

**Exercise 3.** (3 points) We study speed of convergence.

1. (1 point) Give the definition of superlinear and quadratic convergence.
2. (2 points) Show that the sequence  $x_k = 1 + \left(\frac{1}{2}\right)^{2^k}$  converges quadratically to 1.

**Exercise 4.** (7 points) Consider the function  $f(x, y) = 100(y - x^2)^2 + (1 - x)^2$ .

1. (2 points) Compute  $\nabla f$ ,  $\nabla^2 f$ .
2. (2 points) Find the unique global minimizer  $\mathbf{x}^*$  of  $f$ . Justify why it is a global minimizer. Evaluate the gradient and Hessian at  $\mathbf{x}^*$ .
3. (3 points) Write down a steepest descent direction algorithm to minimize  $f$ . At each step, use the backtracking algorithm to find the step length  $\alpha$ .