# Numerical Optimization

## Assignment 1

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## Introduction

This assignment shows testing results of different algorithms' performances by using python.

## Theoretical Exercises

#### Exercise 1

When  $f: \mathbb{R}^d \to \mathbb{R}$  is a strictly convex function, for all  $\alpha \in (0,1)$  holds  $f(\alpha x + (1-\alpha)y) < \alpha f(x) + (1-\alpha)f(y)$ . Then,  $f(Ax) = f(A\alpha x + A(1-\alpha)y) < \alpha f(Ax) + (1-\alpha)f(Ay)$ , so f(Ax) is also a strictly convex function.

Now we know Q is a symmetric positive definite matrix,  $Q = LL^{\mathsf{T}}$ , where L is a lower triangular matrix with unit diagonal elements.  $x^{\mathsf{T}}Qx = x^{\mathsf{T}}LL^{\mathsf{T}}x = ||L^{\mathsf{T}}x||^2$ . Since  $||x||^2$  is a strictly convex function and  $L^{\mathsf{T}}$  is an invertible matrix. by using previous proved statement,  $x^{\mathsf{T}}Qx$  is strictly convex.

#### Exercise 2

$$\begin{split} f_3(x) &= \log(\epsilon + f_1(x)), \epsilon = 10^{-4} \\ f_1(x) &= \sum_{i=1}^d \alpha^{\frac{i-1}{d-1}} x_i^2, \alpha = 1000 \\ \text{When } d &= 1, \, \alpha^{\frac{0}{0}} = 1, \, f_3(x) = \log(\epsilon + x_i^2), \epsilon = 10^{-4} \\ \nabla f_3(x) &= \frac{2x_i}{\epsilon + x_i^2} \\ \nabla^2 f_3(x) &= \frac{2(\epsilon + x_i^2) - (2x_i)^2}{(\epsilon + x_i^2)^2} \\ \text{When } |x| &< \sqrt{\epsilon}, \, \nabla^2 f_3(x) > 0, \, \text{so } f_3(x) \, \text{ is a strictly convex function.} \end{split}$$

#### Exercise 3

$$f_4(x) = \sum_{i=1}^d h(x_i)^2 + 100 \cdot h(-x_i)^2,$$
  
$$h(x) = \frac{\log(1 + \exp(q \cdot x))}{q}, q = 10^4$$

## Conclusion

From this assignment, I had a deeper view on the benchmark functions from previous assignment. Also, used python to test different algorithms' performances in common optimization toolboxes in order to get a visualized and further understanding behind the algorithms.