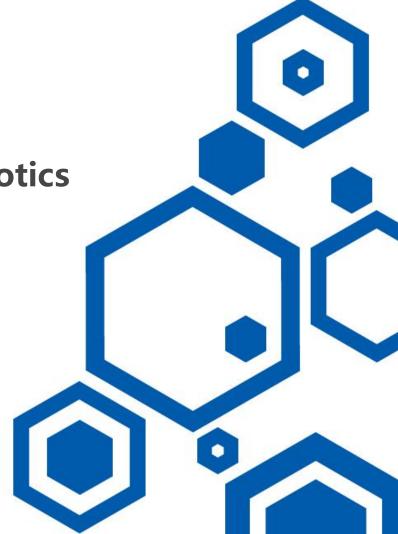


Numerical Optimization in Robotics

L1--作业提示





前言



L1

- 了解不同数值优化问题 需要用什么求解方法
- 可以使用成熟求解器来实现求解

L2

- 满足L1
- 了解不同算法的基本原理与推导过程
- 手动编程实现优化求解 器算法的全过程

L3

- 满足L2
- 可以手动推导已有不同 算法的实现方法
- 针对关键环节给出证明





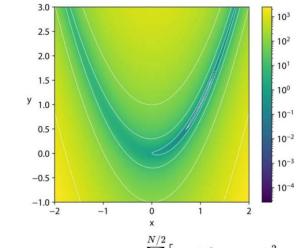




作业题目回顾



1. Problem description:Linear-search Steepest Gradient Descent



Unqualified: The results are incorrect, or the assignment is not written in the required format.

Qualified: The results are somewhat different from the standard results, but there is a correct knowledge and understanding of the assignment requirements.

Good: The results are correct. The program converges correctly to the minimum value of the objective function.

OutStanding: There are two chance for you to get an 'OutStanding':

- (1) Your program can optimize a Rosenbrock function for a given N of dimensions.
- (2) Your program can visualize a two-dimension Rosenbrock function.

What's more, you must have an in-depth understanding of Rosenbrock function optimization, which should be expressed in your report.

 $f(\mathbf{x}) = f(x_1, x_2, \dots, x_N) = \sum_{i=1}^{N/2} \Bigl[100 ig(x_{2i-1}^2 - x_{2i} ig)^2 + (x_{2i-1} - 1)^2 \Bigr]$

You have to implement Linear-search Steepest Gradient Descent to optimize the function given above. It is obvious that the global minimum of the function is $(1,1,...,1)_{N\times 1}$, so you can check if the result of your program is admissible.

算法实现

end



算法 1: Steepest Gradient Descent with Armijo Condition

```
k := 0, c := c_0, \mathbf{x} := \mathbf{x}_0, found := false;
while not found do
      \mathbf{g} := \nabla f(\mathbf{x});
      if ||\mathbf{g}|| < \epsilon then
             found := true;
      else
             \alpha := 1, \mathbf{x}_{new} := \mathbf{x} - \alpha \mathbf{g};
             while f(\mathbf{x}_{new}) > f(\mathbf{x}) - c\alpha \mathbf{g}^{\mathsf{T}} \mathbf{g} \mathbf{do}
                  \alpha := \alpha/2, \mathbf{x}_{new} := \mathbf{x} - \alpha \mathbf{g};
             end
             x := x_{new}, k := k + 1:
      end
```

需要完成工作

- 1)函数f(x) 与其导数g(x)实现
- 2)算法实现

算法实现



函数f(x) 与其导数g(x)实现

$$f(\mathbf{x}) = f(x_1, x_2, \dots, x_N) = \sum_{i=1}^{N/2} \Bigl[100 ig(x_{2i-1}^2 - x_{2i} ig)^2 + (x_{2i-1} - 1)^2 \Bigr]$$

函数 N=2

$$f(x_0, x_1) = 100 * (x_0^2 - x_1)^2 + (x_0 - 1)^2$$

函数的梯度 N=2

$$\frac{\delta f}{\delta x_0} = 400 * x_0 (x_0^2 - x_1) + 2 * (x_0 - 1)$$

$$\frac{\delta f}{\delta x_1} = -200 * (x_0^2 - x_1)$$

Rosenbrock函数

$$f(x) = f(x_1, x_2, ..., x_n)$$

$$= \sum_{i=1}^{N/2} [100(x_{2i-1}^2 - x_{2i})^2 + (x_{2i-1} - 1)^2]$$

$$= [100 * (x_1 - x_2)^2 + (x_2 - 1)^2] + ... + [100 * (x_{N-1} - x_N)^2 + (x_N - 1)^2]$$

修改为从0开始索引

$$f(x) = f(x_1, x_2, ..., x_n)$$

$$= \sum_{i=1}^{N/2} [100(x_{2i-2}^2 - x_{2i-1})^2 + (x_{2i-2} - 1)^2]$$

$$= [100 * (x_0^2 - x_1)^2 + (x_0 - 1)^2] + ... + [100 * (x_{N-2}^2 - x_{N-1})^2 + (x_{N-2} - 1)^2]$$

函数的求导:

$$\frac{\delta f}{\delta x_{2i-2}} = 400 * x_{2i-2} * (x_{2i-2}^2 - x_{2i-1}) + 2 * (x_{2i-1} - 1)$$

$$\frac{\delta f}{\delta x_{2i-1}} = -200 * (x_{2i-2}^2 - x_{2i-1})$$

建议



- 1) 在学习课程与作业的时候,最好能够结合视频课程、推荐书籍以及网络资源等多方面的资源,可以加强自己的理解
- 2) 学习过程中,鼓励实现不同的算法来对比,甚至使用成熟求解器来做对比
- 3) 函数的实现及求导,需要认真仔细,最好能够自己设计一些可控的测试函数进行自测
- 4)作业过程中,可以尝试自己调参,看看过程中不同参数对结果有些什么影响
- 5) 作业的提交,建议按照规范

在线问答







感谢各位聆听 Thanks for Listening

