

# Assignment of Nonlinear Optimization course

Files' description:

- **hessian.m**: calculates hessian matrix.
- **grad.m**: calculates the gradient.
- **rosen.m** : rosenbrock function.
- **fun.m**:  $f(x_1, x_2) = x_1^2 + 4x_2^2 + x_1x_2 - 2x_1 - x_2$
- **armijo.m**: armijo condition.
- **newton\_method.m**: classical newton method.
- **gradient\_descent.m**: classical gradient descent with armijo rule.

## 1) Fist optimization

Function to be minimized:  $f(x_1, x_2) = x_1^2 + 4x_2^2 + x_1x_2 - 2x_1 - x_2$ .

It is a convex function since it's Hessian matrix is positive defined:

$$\nabla^2 f(x_1, x_2) = \begin{bmatrix} 2 & 1 \\ 1 & 8 \end{bmatrix}, \text{ with eigenvalues } \lambda_1 = 5 + \sqrt{10} > 0 \text{ e } \lambda_2 = 5 - \sqrt{10} > 0.$$

1.1) Armijo function finds an acceptable step  $\alpha$  regarding the direction  $d = -\nabla f(1, 1)$ , that satisfies:

$$f(x + \alpha d) \leq f(x) + c\alpha \nabla f(x)^T d \text{ with } c = 0.5.$$

```
armijo(@fun, [1,1])
```

We know that  $f\left(\begin{bmatrix} 1 \\ 1 \end{bmatrix} - \alpha \nabla f(1, 1)\right) = f\left(\begin{bmatrix} 1 \\ 1 \end{bmatrix} - \alpha \begin{bmatrix} 1 \\ 8 \end{bmatrix}\right)$ .

More precisely  $g(\alpha) = (1 - \alpha)^2 + 4(1 - 8\alpha)^2 + (1 - \alpha)(1 - 8\alpha) - 2(1 - \alpha) - (1 - 8\alpha)$ .

Simplyfing,  $g(\alpha) = 265\alpha^2 - 65\alpha + 3$ .

Thus,  $\operatorname{argmin}(g(\alpha)) \approx 0.1226415$ .

Armijo found an acceptable step  $\alpha = 0.0625$ , that is:

$$\begin{aligned} g(0.0625) &< g(0) - (0.5)(0.0625)(9) \\ -0.02734375 &< 3 - 0.28125 \end{aligned}$$

1.2) Optimization using Classical Newton Method, since it is a convex function.

```
newton_method(@fun, [10090,700])
```

```
ans = 1x2
```

```
1.0000 -0.0000
```

### 1.3) Optimization using Gradient Descend with Armijo condition.

```
gradient_descent(@fun, [4,10])
```

```
ans = 1x2  
1.0000 0.0000
```

## 2) Second optimization

Function to be optimized (rosenbrock):  $f(x_1, x_2) = 100(x_2 - x_1^2)^2 + (1 - x_1)^2$ .

It is a non-convex function, so Classical Newton Method cannot solve. Let's show:

$$\nabla^2 f(x_1, x_2) = \begin{bmatrix} -400(x_2 - x_1^2) + 800x_1^2 + 2 & -400x_1 \\ -400x_1 & 200 \end{bmatrix}.$$

So, if  $\nabla^2 f(0, 1) = \begin{bmatrix} -398 & 0 \\ 0 & 200 \end{bmatrix}$ , then it's eigenvalues are  $\lambda_1 = -398$  and  $\lambda_2 = 200$  what proves its non-convexity property.

2.1) Find the minima using Armijo condition with direction  $d = -\nabla f(0, 0.005)$ .

```
armijo(@rosen, [0,0.005])
```

```
ans = 0.0156
```

We know that  $f\left(\begin{bmatrix} 0 \\ 0.005 \end{bmatrix} - \alpha \nabla f(0, 0.005)\right) = f\left(\begin{bmatrix} 0 \\ 0.005 \end{bmatrix} - \alpha \begin{bmatrix} -2 \\ 1 \end{bmatrix}\right)$ .

More precisely  $g(\alpha) = 1600\alpha^4 + 800\alpha^3 + 100\alpha^2 - 5\alpha + 1.0025$ .

Thus,  $\text{argmin}(g(\alpha)) \approx 0.019963$ .

Armijo found an acceptable step  $\alpha = 0.0156$ , that is:

$$\begin{aligned} g(0.0156) &\leq g(0) - (0.5)(0.0156)(5) \\ 0.951968 &\leq 1.0025 - 0.039 (= 0.9635) \end{aligned}$$

2.2) Optimization using Gradient Descent with armijo condition.

```
gradient_descent(@rosen, [4,10])
```

```
ans = 1x2  
1.0000 1.0000
```