

WROCLAW UNIVERSITY OF TECHNOLOGY
DEPARTMENT OF ELECTRONICS

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**Optimization Methods:
Nonlinear Equations**

AUTHOR:
Jaroslaw M. Szumega

SUPERVISOR:
Rafal Zdunek, D.Sc, K-4/W4

GRADE:

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Chapter 1

Solution to the given problems

Problem 1: Show that the following system of nonlinear equations:

$$\begin{aligned}F_1(\mathbf{x}) &= x_1^3 + 3x_1^2 + 3x_1 - x_2 = 0, \\F_2(\mathbf{x}) &= x_1^2 + 2x_1 - x_2 + 1 = 0,\end{aligned}$$

has the global minimum at $\mathbf{x}^* = [0.46557 \quad 2.1479]^T$, the local minimum at $\mathbf{x}_1 = [-1 \quad -0.5]^T$,
and the saddle point at $\mathbf{x}_2 = \left[-\frac{1}{3} \quad -\frac{7}{54}\right]^T$.

In general, the unconstrained nonlinear least squares problem is equivalent to:

$$\min_x \sum_{i=1}^N F_i^2(x)$$

with the objective function:

$$f(x) = \frac{1}{2} \sum_{i=1}^N F_i^2(x)$$

In the case of the following task, objective function will be

$$\begin{aligned}f(x) &= \frac{1}{2}(F_1^2 + F_2^2) \\&= \frac{1}{2}[(x_1^6 + 6x_1^5 + 15x_1^4 - 2x_1^3x_2 + 18x_1^3 - 6x_1^2x_2 + 9x_1^2 - 6x_1x_2 + x_2^2) \\&+ (x_1^4 + 4x_1^3 - 2x_1^2x_2 + 6x_1^2 - 4x_1x_2 + 4x_1 + x_2^2 - 2x_2 + 1)] \\&= \frac{1}{2}(x_1^6 + 6x_1^5 + 16x_1^4 - 2x_1^3x_2 + 22x_1^3 - 8x_1^2x_2 + 15x_1^2 - 10x_1x_2 + 4x_1 + 2x_2^2 - 2x_2 + 1)\end{aligned}$$

And the points to check are:

$$\begin{aligned}\mathbf{x}^* &= [0.46557 \quad 2.1479]^T \\ \mathbf{x}_1 &= [-1 \quad -0.5]^T \\ \mathbf{x}_2 &= \left[-\frac{1}{3} \quad -\frac{7}{54}\right]^T\end{aligned}$$

Firstly, we can calculate the cost function gradient to verify above-mentioned points:

$$\nabla f(x^*) = 0$$

$$\begin{aligned}\frac{\delta f(x)}{\delta x_1} &= 3x_1^5 + 15x_1^4 + 32x_1^3 - 3x_1^2x_2 + 33x_1^2 - 8x_1x_2 + 15x_1 - 5x_2 + 2 \\ \frac{\delta f(x)}{\delta x_2} &= -x_1^3 - 4x_1^2 - 5x_1 + 2x_2 - 1\end{aligned}$$

And in fact, the mentioned in task description points are recognized as a solutions of the cost function gradient.

To check their nature, we need to calculate the Hessian and evaluate it according to the stationary points.

$$\nabla^2 f(x) = \begin{bmatrix} \frac{\delta^2 f(x)}{\delta x_1^2} & \frac{\delta^2 f(x)}{\delta x_1 x_2} \\ \frac{\delta^2 f(x)}{\delta x_1 x_2} & \frac{\delta^2 f(x)}{\delta x_2^2} \end{bmatrix}$$

$$\nabla^2 f(x) = \begin{bmatrix} 15x_1^4 + 60x_1^3 + 96x_1^2 - 6x_1x_2 + 66x_1 - 8x_2 + 15 & -3x_1^2 + 8x_1 + 2 \\ -3x_1^2 + 8x_1 + 5 & 2 \end{bmatrix}$$

$$\text{for point } [0.46557 \ 2.1479] = \begin{bmatrix} 50.11 & -9.37 \\ -9.37 & 2 \end{bmatrix}$$

$\det > 0, M1 > 0, f(x) = 5.01e - 11$, there is a minimum (global)

$$\text{for point } [-1 \ -0.5] = \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix}$$

$\det > 0, M1 > 0, f(x) = 0.25$, there is a minimum

$$\text{for point } [-\frac{1}{3} \ -\frac{7}{54}] = \begin{bmatrix} 2.41 & -2.66 \\ -2.66 & 2 \end{bmatrix}$$

$\det < 0, M1 > 0, f(x) = 0.33$, there is a saddle point

To make all the calculations above, the following script in Python was written to perform symbolic computations:

```

1 #!/usr/bin/python
2
3 from sympy import diff, Symbol, latex, Matrix
4
5
6 def optimization(function, x1, x2):
7
8     print latex(function)
9
10    #calculating the gradient's elements
11    firstx1 = diff(function, x1)
12    firstx2 = diff(function, x2)
13
14    print firstx1.expand()
15    print firstx2.expand()
16
17
18    #calculations for Hessian elements
19    secondx1x1 = diff(function, x1, x1)
20    secondx1x2 = diff(function, x1, x2)
21    secondx2x1 = diff(function, x2, x1)
22    secondx2x2 = diff(function, x2, x2)
23
24    print "\n Hessian elements:\n\n"
25    print latex(secondx1x1) + "\t" + latex(secondx1x2)
26    print latex(secondx2x1) + "\t" + latex(secondx2x2)
27
28
29    point1 = [0.46557, 2.1479]
30    point2 = [-1.0, -0.5]
31    point3 = [-1.0/3, -7.0/54]
32
33    point = point3
34    print secondx1x1.subs(x1, point[0]).subs(x2, point[1]);
35    print secondx1x2.subs(x1, point[0]).subs(x2, point[1]);
36    print secondx2x1.subs(x1, point[0]).subs(x2, point[1]);
37    print secondx2x2.subs(x1, point[0]).subs(x2, point[1]);
38    print "\n\n"
39    print function.subs(x1, point[0]).subs(x2, point[1]);
40
41
42 def main():
43     x1 = Symbol('x1')
44     x2 = Symbol('x2')
45
46     f1 = x1**3 + 3*x1**2 + 3*x1 - x2
47     f2 = x1**2+2*x1 -x2 + 1
48     f = (f1**2 + f2**2)/2
49
50     function = f.expand()
51     optimization(function, x1,x2)
52
53 main()

```

Problem 2: Solve the following system of nonlinear equations with the selected nonlinear least squares methods.

$$F_1(\mathbf{x}) = x_1^3 - 3x_1^2 + 3x_1 - x_2 - 3 = 0,$$

$$F_2(\mathbf{x}) = x_1^2 - 2x_1 - x_2 = 0.$$

Draw the contour lines of the cost function in the nonlinear least square problem. Check the optimality points.

Chapter 2

Listings of algorithms

2.1 Coded selected algorithms

Algorithm 1 - Simplex algorithm

Algorithm 2 - Revised Simplex algorithm

Bibliography

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