Numerical Optimization Solution to exercise sheet

review on 18.12.2024 during the exercise class

1. (Null Space Method)

Let $A \in \mathbb{R}^{n \times n}$ be a symmetric and positive semidefinite matrix, $B \in \mathbb{R}^{m \times n}$ and $g \in \mathbb{R}^m$ with $m \le n$. Consider the following equality constrained optimization problem

minimize
$$f(x) = \frac{1}{2}x^T A x + a^T x$$
 subject to $Bx = b$. (1)

a) Let $Z \in \mathbb{R}^{n \times (n-m)}$ be a null space matrix of B, that means that the columns of Z are a basis for the null space of the matrix B. Prove: If rg(B) = m and $Z^T A Z$ is positive definite, then the matrix

$$K = \begin{pmatrix} A & B^T \\ B & 0 \end{pmatrix}$$

is invertible. Especially there exists a unique pair $(x^*, \lambda^*) \in \mathbb{R}^n \times \mathbb{R}^m$, which is the solution of the saddlepoint problem

$$\begin{pmatrix} A & B^T \\ B & 0 \end{pmatrix} \begin{pmatrix} x^* \\ \lambda^* \end{pmatrix} = \begin{pmatrix} -a \\ b \end{pmatrix}. \tag{2}$$

b) Let the assumption of subtask a) be fullfilled. Prove that the unique solution of (2) is the unique solution of (1).

(4 + 4 = 8 Points)

Solution:

a) Let $(x, \lambda) \in \mathbb{R}^n \times \mathbb{R}^m$ be arbitrary such that:

$$\begin{pmatrix} A & B^T \\ B & 0 \end{pmatrix} \begin{pmatrix} x \\ \lambda \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}. \tag{3}$$

From the second equation we deduce Bx=0, from which $x\in \ker(B)$ follows. Therefore, there exists a vector $w\in \mathbb{R}^{n-m}$ with x=Zw, because the columns of Z are a basis of $\ker(B)$. From Bx=0 it follows also $x^TB^T=0$. Therefore

$$0 = \begin{pmatrix} x \\ \lambda \end{pmatrix}^T \begin{pmatrix} A & B^T \\ B & 0 \end{pmatrix} \begin{pmatrix} x \\ \lambda \end{pmatrix} = x^T A x = w^T Z^T A Z w.$$

Due to the assumption that Z^TAZ is positive definite, w=0 follows directly. From that we can deduce x=Zw=0. It is left to prove that $\lambda=0$. From the first equality of (3) it follows with x=0 that $B^T\lambda=0$. Due to the full rank of B and therefore its surjectivity, we have that B^T is injective and so $\lambda=0$ must hold.

- b) The function f is convex, because the Hessian $\nabla^2 f = A$ is positive semi-definite. The constraints are affine linear, so that the optimization problem is convex and every solution of the KKT-system is a global solution of the optimization problem. The claim follows immediately, because the KKT system is given in (2).
- 2. (Null Space Method, (MATLAB))
 Consider the linear quadratic equality constrained optimization problem (1).
 - a) Apply the Matlab function nullspace_method.m (given in the material) to the equality constrained optimization problem (1) using

$$A := \begin{pmatrix} 1 & -1 & 0 & 0 & 0 \\ -1 & 2 & 1 & 0 & 0 \\ 0 & 1 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix}, \quad a := \begin{pmatrix} 0 \\ -2 \\ -2 \\ -1 \\ -1 \end{pmatrix}, \quad B := \begin{pmatrix} 1 & 3 & 0 & 0 & 0 \\ 0 & 0 & 1 & 1 & -2 \\ 0 & 1 & 0 & 0 & -1 \end{pmatrix}, \quad b := \begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \end{pmatrix}.$$

- b) Adjust the method nullspace_method.m to nonlinear_nullspace_method.m such that general nonlinear functions $f: \mathbb{R}^n \to \mathbb{R}$ can be optimized.
- c) Apply the method nonlinear_nullspace_method.m to the minimization problem given in a) by using the initial value $z^{(0)} = (1,1)^T$.

(6 Points)

Solution:

a) The Matlab-function could look like

```
clear, close all
clc
% Initializiation
B=[1 \ 3 \ 0 \ 0 \ 0; \ 0 \ 0 \ 1 \ 1 \ -2; \ 0 \ 1 \ 0 \ 0 \ -1];
b = [0; 0; 0];
a = [0; -2; -2; -1; -1];
A=[1 -1 0 0 0; -1 2 1 0 0; 0 1 1 0 0; 0 0 0 1 0; 0 0 0 0 1];
z0 = ones(2,1);
% Define the nonlinear function
f = 0(x) 1/2*x**A*x + a**x;
% Anwendung des Nullraum-Verfahrens
[x, lambda] = nullspace_method(A, a, B, b);
[x_nonlin] = nonlinear_nullspace_method(f, z0, B, b);
% Berechnung des Fehlers in der euklidischen Norm
x_{exact} = 1/43 * [-33; 11; 27; -5; 11];
error = norm(x - x_exact);
display(['Fehler zwischen exaktem und numerischem Ergebnis: ' num2str(error)]);
error = norm(x_nonlin - x_exact);
display(['Fehler zwischen exaktem und numerischem Ergebnis: ' num2str(error)]);
```

b) The Matlab-function could look like

```
function [x] = nonlinear_nullspace_method(f, z0, B, b)
```

```
n=size(B,2);
m=size(B,1);

%-- (1) QR-decomposition and splitting
[Q, R]=qr(B');
R = R(1:m,1:m);
Y = Q(1:n,1:m);
Z = Q(1:n,m+1:n);

%-- (2) determine x_Y
x_Y = (R')\b;

%-- (3) determine x_Z by minimizing f(Y x_y + Z z)
w = Y*x_Y;
z = fminunc(@(z)f(w + Z*z), z0, optimset('Display','off'));

% determine the full solution
x = w + Z*z;
end
```

3. (Hanging chain, Matlab)

We model a chain which is fixed on both ends. To get a mathematical model we consider the chain as point masses glued together by springs. This is depicted in the following:

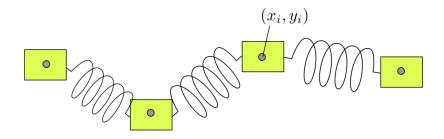


Figure 1: Chain modeled as point masses and springs.

The total energy of the system is given by $E_{\text{tot}} = E_{\text{pot}} + E_{\text{spring}}$, where $E_{\text{pot}} = mgh$ is the potential energy of the mass m with gravity $g = 9.81m/s^2$ at a height of h and $E_{\text{spring}} = \frac{1}{2}D\left((x_i - x_{i+1})^2 + (y_i - y_{i+1})^2\right)$ is the energy of a spring with spring constant D tensioned between the points (x_i, y_i) and (x_{i+1}, y_{i+1}) . Now, nature tries to find the steady state of the system by minimizing the total energy. This means numerically, we face the problem

$$\min_{x,y} E_{\text{tot}}$$
s.t. $h(x,y) = 0$,

where h models where the chain is fixed. Implement and solve this problem with MATLAB. Play around with h, D and m. Plot the resulted chain.

(8 Points)

Solution: We can extend everything easily to the 3D case and decorate a christmas tree with a light chain. The MATLAB-function could look like

```
clear, close all
```

```
clc
 % Source:
 % https://de.mathworks.com/matlabcentral/fileexchange/9337-xmas-tree
ov_xmasTree();
% points for the light chain
% coordinates on the tree
x_{\text{tree}} = [-0.1153, 2.986, 3.059, -4.7118, -5.0628, 5.01336, ...
           5.15578, -3.3961, -7.445, -4.9825, 5.1589];
y-tree = [-2.7033, -0.9761, 3.657, 2.729, -3.4359, -4.31467, ...
           5.4282, 7.1485, -2.6252, -7.38934, -6.8566];
z_{\text{tree}} = [22.657, 21.722, 18.024, 17.241, 15.1467, 13.739, ...
          11.9194, 10.994, 10.7183, 9.6119, 8.6214];
% size of equality constraints (actually a third of it)
M = size(x_tree, 2);
% calculate how many lights between points
length_inbetween = zeros(1,M-1);
for i = 1:M-1
    length_inbetween(i) = sqrt((x_tree(i)-x_tree(i+1))^2 ...
                                + (y_tree(i)-y_tree(i+1))^2 ...
                                + (z_tree(i)-z_tree(i+1))^2);
total_length = sum(length_inbetween);
N = 91;
n = 3 * (N+1);
mass\_vec = 20*ones(N+1,1);
C_vec = 10*N*ones(N,1);
        = 9.81;
         = @(x) total_energy(x, mass_vec, C_vec, g);
% constraints
% calculate the indices of constraints (this determines how many lights
% we want to have inbetween)
index_constr = floor([0, length_inbetween]/total_length * N);
for i = 2:size(index_constr,2)
    index_constr(i) = index_constr(i-1) + index_constr(i);
end
index_constr(1)
                    = 1;
index_constr(end)
                  = N;
index_constr(end-1) = index_constr(end-1) + 5;
index_constr
                    = cast(index_constr, "int32");
        = 3 * M;
B
         = zeros(m,n);
        = [x_tree, y_tree, z_tree]';
% fill in the constraint matrix
for i = 1:M
    % x coordinate
    B(i, index\_constr(i)) = 1;
    % y coordinate
   B(M + i, N + 1 + index\_constr(i)) = 1;
    % z coordinate
    B(2*M + i, 2*N + 2 + index_constr(i)) = 1;
end
% initial value (not needed with the other algorithm)
```

The results are given in the following figure



Figure 2: A christmas tree with decorated with a light chain, obtained by the quadratic optimization algorithm nullspace method. The white chain goes around the christmas tree.

Merry Christmas and a Happy New Year