



**Silesian University
of Technology**

FACULTY OF AUTOMATIC CONTROL, ELECTRONICS
AND COMPUTER SCIENCE

Advanced Optimization Methods

Constrained dynamic optimization (penalty methods)

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1 Task

The task was to determine optimal control for a system described by state equation, with given x_0 and constraints. The equations are specified in the instruction in problem number 3. Selected x_6 parameter was $x_6 = 6$. In accordance with Professor's e-mail, the constraint $x_3 = 5$ was removed.

2 Hand-written part of solution

Task: $x_{i+1} = 2x_i + u_i$ constraints: $x_6 = 6$
 $J = \sum_{i=0}^{i=5} (x_i^2 + 2u_i^2)$ $-1 \leq u_i \leq 3 \Leftrightarrow u_i \leq 3 \wedge -u_i \leq 1$
 $N=6$
 $a = [6; -1; 3]$
 $u^0 = [1; 2; 3; 4; 5; 6]$

$$J_v = J + \left\{ 0.5 \|x_6 - v_1\|^2 + \frac{1}{2} \sum_{n=0}^5 \left[\sum_{j=1}^1 [(-u_{nj} + 1) \cdot \max(0, -u_{nj} + 1)] + \sum_{j=1}^1 (u_{nj} - 3) \cdot \max(0, u_{nj} - 3) \right] \right\}$$

 $\textcircled{2} \alpha = 0,1$ $\textcircled{3} v^0 = a = [6; -1; 3]$
 $\beta = 5$
 $\varepsilon_1 = 0,1$ $\textcircled{4} u^0 = [-1,2965; -0,8429; -0,2804; 0,0717; 0,4777; 0]$
 $\varepsilon_2 = 0,1$
 $c = 2$ $x^0 = [1; 0,7035; 0,5611; 0,8479; 1,7675; 4,0127]$
 $t = 5$
 $\textcircled{5} r^0 = \begin{bmatrix} x(a) - x_6 \\ \max(\max(0, u^0 - 1)) \\ \max(\max(0, u^0 - 3)) \end{bmatrix} = \begin{bmatrix} -1,9873 \\ 0,2965 \\ 0 \end{bmatrix}$
 $\textcircled{6} \chi = \|v^0 + r^0 - a\| = \left\| \begin{bmatrix} -1,9873 \\ 0,2965 \\ 0 \end{bmatrix} \right\| = 2,0033$
 $\textcircled{7} \chi \leq \varepsilon \text{ false}$
 $\textcircled{8} \chi < c \text{ false}$
 $\textcircled{9} \chi > c^0 \rightarrow t^1 = 5:5 = 25; v^1 = [6,3975; -1,0593; 3]$

Figure 1: Hand written part of solutions with selected parameters

Selected initial values of control vector: $u_0 = 1, u_1 = 2, u_2 = 3, u_3 = 4, u_4 = 5, u_5 = 6$

3 Results

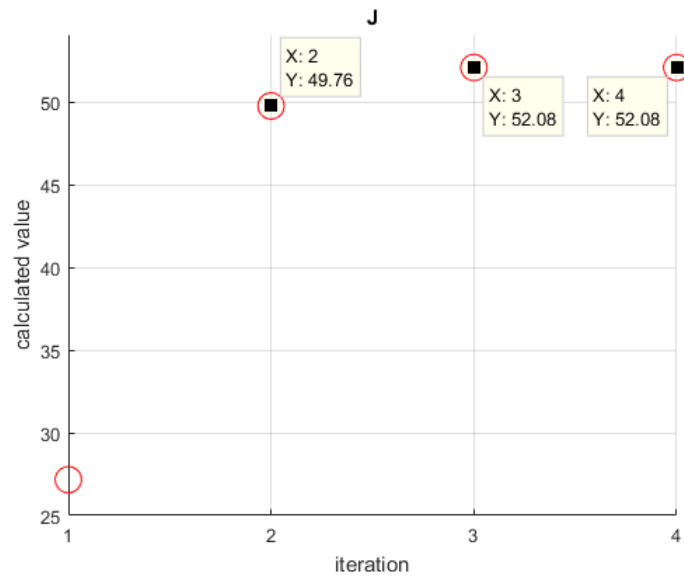


Figure 2: Results obtained for performance index J

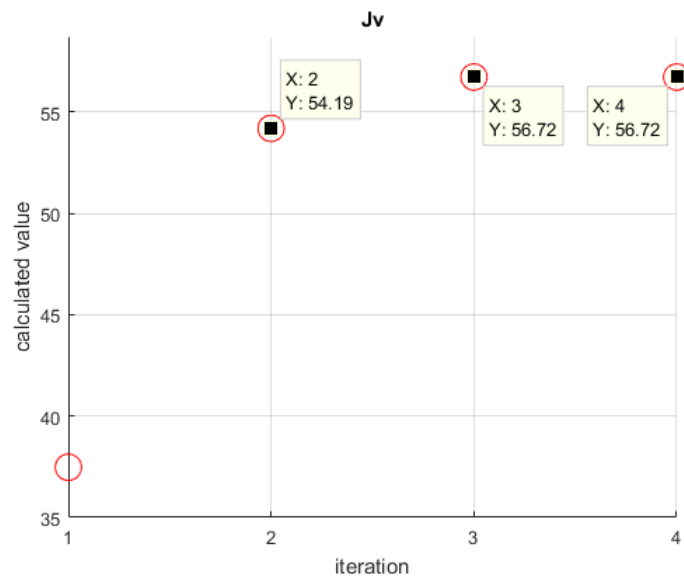


Figure 3: Results obtained for penalty functional

```

1 X values, x(1): 1, x(2): 0.82318, x(3): 0.74425, x(4): 1.2235, x(5): 2.6204, x(6): 5.9826
2 U values, u(1): -1.1768, u(2): -0.90211, u(3): -0.26497, u(4): 0.17336, u(5): 0.74178, ...
   u(6): -1.9861e-05

```

4 Conclusions

The algorithm as usual needed small amount of iterations to achieve the solution. As can be seen the constraints are not fully satisfied, the u_1 still violates them therefore the value of penalty functional is still greater than the function being minimized. The algorithm has stopped after meeting both stop criteria - achieving absolute value of consecutive penalty functionals being subtracted smaller than given accuracy and calculated gamma also being smaller than given accuracy. As expected, the value of found x_6 is within given precision, which was set to 0.1.

5 Source code

main.m

```
1  clc
2  clear all
3  %%%%%
4  x_1 = 1;
5  x_6 = 6;
6  n=6;
7  %%%%% defining parameters and initial values%%%%
8  alfa = 0.1;
9  beta = 5;
10 epsilon = 0.1;
11 epsilon_2 = 0.1;
12 c = 2;
13 t = 5;
14 a = [x_6, -1, 3]; %values of constraints, x_6=7, ...
    -1<u_i<3
15 v = a; %step 3
16 u0 = [1,2,3,4,5,6] %random values of initial control
17 u=u0
18 i = 1;
19
20 while true
21     Jv_function = @(u)calculate_penalty(x_1, u, n, v, t); %step 1
22     u = fminsearch(Jv_function, u); %finding un, step 4
23     [Jv(i), x, J(i)] = calculate_penalty(x_1, u, n, v, t); %finding xn, step 4
24     r = calculate_r(x(6), v, u); %step 5
25     gamma = norm(v+r-a); %step 6
26
27     if gamma > epsilon %step 7
28         if gamma < c %step 8
29             v= a-r;
30             c = alfa*beta;
31         else %step 9
32             t = beta*t;
33             v = a - r / beta;
34         end
35     end
36     gamma = norm(v+r-a);
37
38     if (i >1) %step 10
39         j_jv = abs(prev_Jv - Jv(i))
40         if gamma <= epsilon && j_jv < epsilon_2
41             break
42         end
43     end
44
45     prev_Jv=Jv(i);
46     i=i+1;
```

```

47 end
48
49 visualize_results(x,u,J,Jv)

```

calculate_penalty.m

```

1 function [Jv, x, J] = calculate_penalty(x0 , u, n, v, t)
2 x = zeros (1, n);
3 x(1) = x0;
4 for i = 2:n
5     x(i) = 2*x(i-1) + u(i-1);
6 end
7 J = sum(x.^2 + 2*u.^2);
8 R_1 = 0.5 * norm(x(6) - v(1))^2; %v(1) corresponds to x_6, etc
9 R_2 = sum(sum((-u + v(2)).* max(0, -u + v(2)))) ;
10 R_3 = sum(sum((u - v(3)).* max(0, u - v(3)))) ;
11 Jv = J + t*(R_1 + R_2 + R_3);
12 end

```

calculate_r.m

```

1 function r = calculate_r(x6,v,u)
2     r = [0,0,0];
3     r(1) = x6 - v(1); %for constraint x6 = 1
4     r(2) = max(max(0, -u + v(2))); %for u_i ≤ 14.5
5     r(3) = max(max(0, u - v(3))); %for u_i ≥ 14.5
6 end

```

visualize_results.m

```

1 function r = visualize_results(x, u, J, Jv)
2 display_x = ['X values, x(1): ', num2str(x(1)), ...
3             ', x(2): ', num2str(x(2)), ...
4             ', x(3): ', num2str(x(3)), ...
5             ', x(4): ', num2str(x(4)), ...
6             ', x(5): ', num2str(x(5)), ...
7             ', x(6): ', num2str(x(6))];
8
9 display_u = ['U values, u(1): ', num2str(u(1)), ...
10            ', u(2): ', num2str(u(2)), ...
11            ', u(3): ', num2str(u(3)), ...
12            ', u(4): ', num2str(u(4)), ...
13            ', u(5): ', num2str(u(5)), ...
14            ', u(6): ', num2str(u(6))];
15 disp(display_x)
16 disp(display_u)
17 x_vector = [1:length(Jv)];
18 figure
19 scatter(x_vector, Jv, 200, 'red')
20 grid on
21 title("Jv")
22 xlabel("iteration")
23 ylabel("calculated value")
24 ax = gca;
25 ax.XTick = x_vector;
26 ax.YLim(2) = max(Jv)+2;
27
28 figure
29 scatter(x_vector, J, 200, 'red')
30 grid on
31 title("J")

```

```
32 xlabel("iteration")
33 ylabel("calculated value")
34 ax = gca;
35 ax.XTick = x_vector;
36 ax.YLim(2) = max(J)+2;
37 end
```