

1 Exercise PP. 9

Consider the following optimization problem:

$$\begin{aligned} & \text{minimize: } x_1^4 - 2x_1^2x_2 + x_1^2 + x_1x_2^2 - 2x_1 + 4 \\ & \text{subject to: } x_1^2 + x_2^2 - 2 = 0 \\ & \quad 0.25x_1^2 + 0.75x_2^2 - 1 \leq 0 \\ & \quad 0 \leq x_1 \leq 4 \\ & \quad 0 \leq x_2 \leq 4 \end{aligned}$$

Solve the constrained minimization using the Steepest Decent Method.

1.1 MATLAB Files

PP9_main.m

```
1 clear
2 clc
3 close(figure(1))
4
5 PP9_data
6
7 % Open Contour Plot Figure
8 figure(1)
9 fcontour(fx1x2,[0 5 0 5])
10 xlabel('$x_1$', 'Interpreter', 'latex')
11 ylabel('$x_2$', 'Interpreter', 'latex')
12 title('$f(x_1,x_2) = x_1^4 - 2x_1^2 x_2 + x_1^2 +x_1 x_2^2 -2$
    $x_1 +4$', ...
13     'Interpreter', 'latex')
14 dim =[0.15, 0.8, 0.1, 0.1];
15 const_str = {'Constraints:', '$x_1^2+x_2^2-2=0$', ' ', ...
16     '$0.25x_1^2+0.75x_2^2-1 \leq 0$', ' ', '$0 \leq x_1$
17     $\leq 4$',...
18     ' ', '$0 \leq x_2 \leq 4$'};
19 annotation('textbox',dim,'String', const_str, 'Interpreter', '
20     latex', 'BackgroundColor', 'w')
21
22 grid on
23 hold on
24
25 epsilon_min = 0.0001;
26 epsilon = 0.1;
27 miu = zeros(length(hx1x2(X(1),X(2))),1);
28 lambda = zeros(length(gx1x2(X(1),X(2))),1);
```

```
27 % Search Cycle
28 k=0;
29 while k<kmax
30     k=k+1;
31     [Lx1x2,gradL] = auglag(X,f,g,gx1x2,h,epsilon,miu,lambda);
32     [X,X_old] = minimize(X,Lx1x2,gradL,lb,ub,tmax);
33
34
35     % Plot Search Path
36     plot([X_old(1) X(1)],[X_old(2) X(2)],'o-r')
37
38     % KKT Conditions
39     [KKT, KKT_norm,flag]=KKT_fun(X,lambda,miu,grad_f,grad_h,
40         grad_g,gx1x2);
41
42     if flag
43         break
44     end
45
46     [miu,lambda,epsilon] = update(X,hx1x2,gx1x2,miu,lambda,
47         epsilon,epsilon_min);
48
49 end
50 f_obj = fx1x2(X(1),X(2));
51
52 % Plot point of minima in a different color
53 plot(X(1),X(2),'ok',MarkerFaceColor='k')
54
55 % Print the results
56 fprintf([ ...
57     'Number of Iterations: %d\n\n', ...
58     'Point of Minima: [%.4f , %.4f]\n\n', ...
59     'Objective Function Minimum Value after Optimization:
60     %.4f\n\n'], ...
61     k,X(1),X(2), f_obj)
```

PP5_data.m

```
1 % Objective function
2 syms x1 x2
3 f(x1,x2) = x1^4 - 2*x1^2*x2 + x1^2 + x1*x2^2 - 2*x1 + 4;
4 df = gradient(f,[x1,x2]);
5 fx1x2 = matlabFunction(f);
6 grad_f = matlabFunction(df);
7
```

```
8 % Initial point (k=0)
9 X = [3;2];
10 f_obj = fx1x2(X(1),X(2));
11
12 % Maximum number of iterations
13 kmax = 2000;
14 tmax = 2000;
15
16 % Constraints
17 h = x1^2 + x2^2-2;
18 g = 0.25*x1^2 + 0.75*x2^2-1;
19 hx1x2 = matlabFunction(h);
20 grad_h = matlabFunction(gradient(hx1x2, [x1 x2]));
21 gx1x2 = matlabFunction(g);
22 grad_g = matlabFunction(gradient(gx1x2, [x1 x2]));
23 lb=[0; 0];
24 ub=[4; 4];
```

auglag.m

```
1 function [Lx1x2,gradL] = auglag(X,f,g,gx1x2,h,epsilon,miu,
   lambda)
2
3 syms x1 x2
4 g_aux = gx1x2(X(1),X(2));
5 aux = zeros(length(gx1x2(X(1),X(2))),1);
6 aux = sym(aux);
7
8 for i=1:length(gx1x2(X(1),X(2)))
9     if (-epsilon/2)*lambda(i)>g_aux(i)
10         aux(i)=(-epsilon/2)*lambda(i);
11     else
12         aux(i)=g(i);
13     end
14 end
15
16 L(x1,x2) = f + miu'*h + lambda'*sym(aux) + (1/epsilon)*norm(h)
   ^2 + (1/epsilon)*(norm(aux))^2;
17 dL = gradient(L,[x1,x2]);
18
19 Lx1x2 = matlabFunction(L);
20 gradL = matlabFunction(dL);
```

minimize.m

```
1 function [X,X_old] = minimize(X,Lx1x2,gradL,lb,ub,tmax)
2 % Minimizes the function at hand using the Steepest Descent
   Method and the Armijo Linear Search
3
4 t=0;
5 X_old = X;
6 while t<tmax && norm(gradL(X(1),X(2)))>10^-6
7     t=t+1;
8
9     % Search Direction
10    d = -gradL(X(1),X(2));
11
12    c = 0.01;
13    gama = 0.01;
14    delta = 0.5;
15
16    % Define initial step value
17    alfa = c*abs(gradL(X(1),X(2))'*d)/(norm(d)^2);
18    X_new = X+alfa*d;
19
20
21    while any(lb > X_new) || any(ub < X_new)
22        alfa = alfa*delta;
23        X_new = X+alfa*d;
24    end
25
26    % Find optimal step
27    while Lx1x2(X_new(1),X_new(2)) > Lx1x2(X(1),X(2)) + gama*
        alfa*gradL(X(1),X(2))'*d && all(lb < X_new) && all(X_new
        < ub)
28        alfa = alfa*delta;
29        X_new = X+alfa*d;
30    end
31
32    X = X+alfa*d;
33 end
```

KKT_fun.m

```
1 function [KKT, KKT_norm,flag]=KKT_fun(X,lambda,miu,grad_f,
   grad_h,grad_g,gx1x2)
2 % Karush-Kuhn-Tucker Conditions
```

```
3
4 flag = false;
5 KKT = grad_f(X(1),X(2)) + miu.*grad_h(X(1),X(2)) + lambda.*
    grad_g(X(1),X(2));
6 KKT_norm = norm(KKT);
7
8 if KKT_norm<1e-06 && lambda'*gx1x2(X(1),X(2))<1e-06
9     flag = true;
```

update.m

```
1 function [miu,lambda,epsilon] = update(X,hx1x2,gx1x2,miu,lambda
    ,epsilon,epsilon_min)
2
3 miu = miu+2/epsilon*hx1x2(X(1),X(2));
4 lambda = max([zeros(length(gx1x2(X(1),X(2))),1), lambda + (2/
    epsilon)*gx1x2(X(1),X(2))],[],2);
5 ro = 0.95;
6 epsilon_old = epsilon;
7 epsilon = epsilon*ro;
8
9 if epsilon < epsilon_min
10     epsilon = epsilon_old;
11 end
```

1.2 Results

Number of Iterations: 16

Point of Minima: [1.0000 , 1.0000]

Objective Function Minimum Value after Optimization: 3.0000

