

MATH693a
Dr. Peter Blomgren
HW03
Trust Region Methods

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Introduction

In this assignment we were given the objective function:

$$f(x) = 10(x_2 - x_1^2)^2 + (1 - x_1)^2 \quad (1)$$

and asked to draw the contour lines of the quadratic model for it, assuming that for the Hessian of $f(x)$ would hold:

$$B_k = \nabla^2 f(\mathbf{x}_k) \quad (2)$$

and the family of solutions for the Trust Region subproblem.

Part 1

In this section we will use $x_0 = [0, -1]$ as our starting point, thus as the center of the trust region. We display the results of the iterations for some values of the trust region radius D_k , between 0 and 2. We performed Cholesky Factorization on the Hessian to find the values of λ (the diagonal elements of the Hessian) to use in order to shift the 2×2 matrix away from regions of non-definiteness (i.e. with any diagonal elements less than 0).

The table below shows the results in terms of number of iterations and the elapsed time necessary to converge to the optimal value for λ , given a maximum number of iterations fixed at 100 thousands.

D_k	Iterations	Elapsed Time (sec)	Optimal λ
0.5	13778	0.248767	20.0833
1	7201	0.146302	0.0227
1.5	100000	1.659368	19.7345
2	100000	1.721085	1.0984

We can immediately appreciate that for a trust-region radius $D_k \leq 1$, we have a relatively fast convergence to an optimal value of λ . Whereas, for $D_k > 1$, we do not have convergence for the point $x_0 = [0, -1]$. Figure 1 below shows the contour of the model for such point, with the trust region and the step direction.

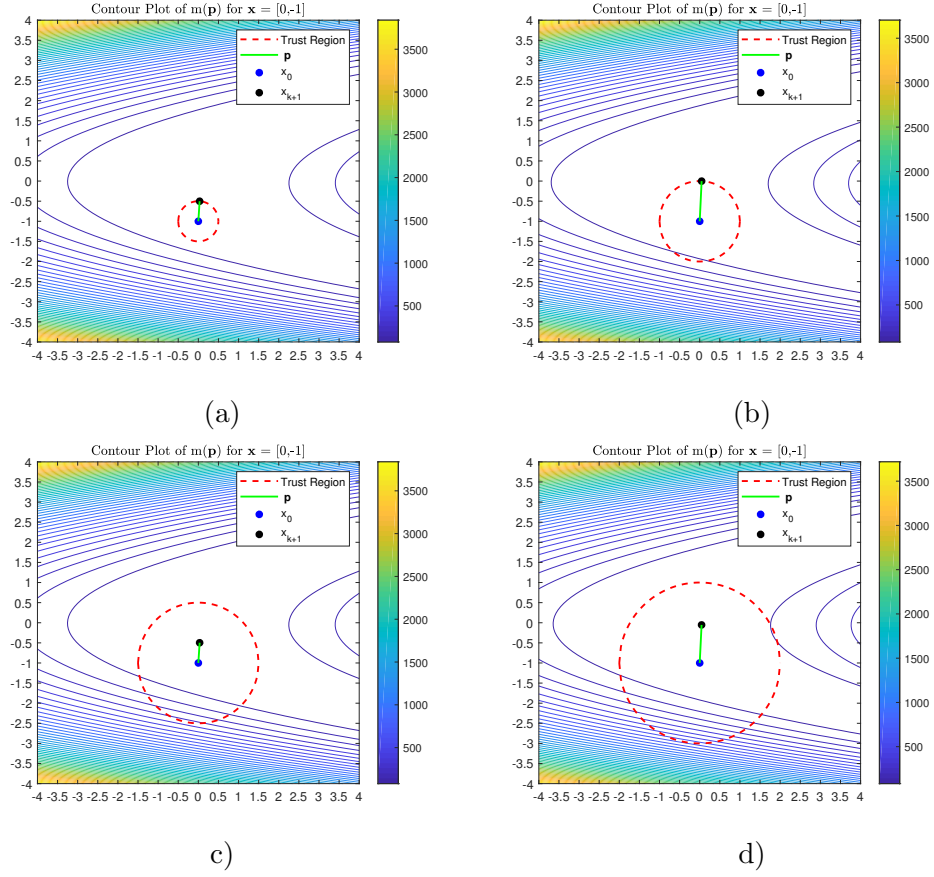


Figure 1: Trust Regions for different values of the radius D_k for the starting point $x_0 = [0, -1]$: a) $D_k = 0.5$, b) $D_k = 1$, c) $D_k = 1.5$, d) $D_k = 2$

We see that in Fig1a) and b), our lambda gives a step direction from the center of the trust region to its boundary, whereas in Fig1c) and d) we see that we stay well inside the region and the step length is much shorter than the trust region radius.

Part 2

In this section we will use $x_1 = [0, 0.5]$ as our starting point, thus as the center of the trust region. We display the results of the iterations for some values of the trust region radius D_k , between 0 and 2. We performed Cholesky Factorization on the Hessian to find the values of λ (the diagonal elements of the Hessian) to use in order to shift the 2×2 matrix away from regions of non-definiteness (i.e. with any diagonal elements less than 0).

The table below shows the results in terms of number of iterations and the elapsed time necessary to converge to the optimal value for λ , given a maximum number of iterations fixed at 100 thousands.

D_k	Iterations	Elapsed Time (sec)	Optimal λ
0.5	89318	1.440988	22.5324
1	23888	0.409682	20.0654
1.5	11193	0.230730	19.3529
2	6537	0.138706	19.0083

We can immediately appreciate that, in this case, for the four trust-region radii D_k s displayed, we have a relatively fast convergence to an optimal value of λ , with running time, number of iterations and λ value that decrease as D_k increases.

Below we plot the contour of the model $m(\mathbf{p})$ for the given point, as well as the trust region and the step direction.

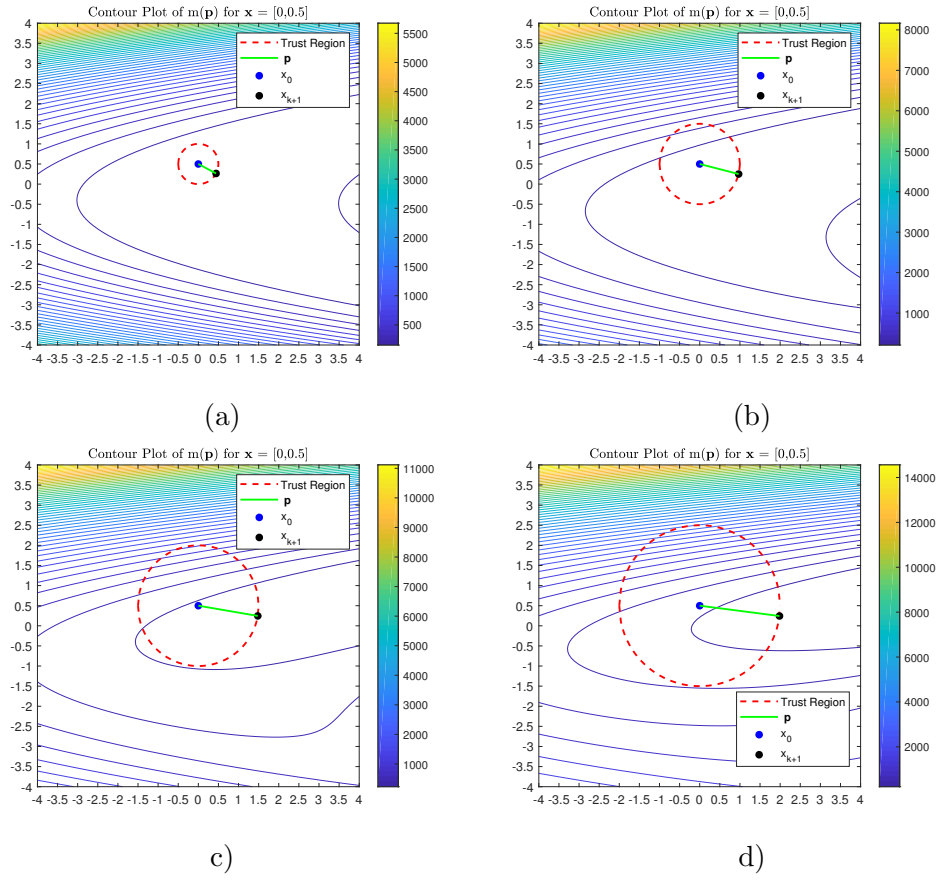


Figure 2: Trust Regions for different various of the radius D_k for the starting point $\mathbf{x}_1 = [0, 0.5]$: a) $D_k = 0.5$, b) $D_k = 1$, c) $D_k = 1.5$, d) $D_k = 2$

Appendix - Matlab Code

```
1 %trust region
2 clear all
3 clc
4 nn = 100000;
5 mu = (min(diag(zeros(2)))));
6 x = [0, -1];
7 Dk = 2;
8 tol = 1e-8;
9 xrange = linspace(-Dk,Dk,101);
10 tic
11 t = false;
12 for ii = 1:nn
13     nabla = grad_f(x);
14     if ii==1
15         A = hess_f(x);
16         mu(ii) = min(abs(diag(A)));
17         if A(ii,ii) < 0
18             mu(ii) = min(abs(diag(A))) + 0.1;
19         end
20     end
21     A = hess_f(x) + mu(ii)*eye(length(hess_f(x)));
22     if ii==2 && (A(ii,ii) < 0)
23         mu(ii) = min(abs(diag(A))) + 0.1;
24     end
25     if mu(ii) < 0
26         mu(ii) = min(abs(diag(A))) + 0.1;
27     end
28     L = chol(x,A);
29     LT = (L)';
30     LI = (L)^-1;
31     q = ((LT)^-1)*(-nabla)';
32     p = (LI)*q;
33     mu(ii+1) = mu(ii) + (((norm(p))/(norm(q)))^2) * ((
        norm(p))-Dk)/Dk;
34     if abs(mu(ii+1)-mu(ii)) < tol
35         t = true;
36         break
37     end
38     mu(ii) = mu(ii+1);
39     xkp1 = x + p';
40 end
```

```

41 toc
42 mu(ii)
43 xcontour = linspace(-4,4,101);
44 [X,Y] = meshgrid(xcontour,xcontour);
45 pt = p';
46 m = zeros(length(xcontour),length(xcontour));
47
48 for jj=1:length(xcontour)
49     x2 = xcontour;
50     for kk=1:length(xcontour)
51         f = h([x2(jj),x2(kk)]);
52         g = grad_f([x2(jj),x2(kk)]);
53         Bk = hess_f([x2(jj),x2(kk)])+mu(kk)*eye(length
                    (hess_f([x2(jj),x2(kk)])));
54         m(jj, kk) = f + pt*(g')+.5*pt*(Bk)*p;
55     end
56 end
57
58 y1 = sqrt(Dk^2-(xrange-x(1)).^2)+x(2);
59 y2 = -sqrt(Dk^2-(xrange-x(1)).^2)+x(2);
60 figure(1)
61 mp = contour(X,Y,m,50);
62 hold on
63 circ1 = plot(xrange,y1,'r—','linewidth',1.5);
64 hold on
65 circ2 = plot(xrange,y2,'r—','linewidth',1.5);
66 hold on
67 x0 = plot(x(1),x(2),'b*','linewidth',2);
68 hold on
69 x1 = plot(xkp1(1),xkp1(2),'k*','linewidth',2);
70 hold on
71 p = plot([x(1) xkp1(1)],[x(2) xkp1(2)],'g-','linewidth
        ',1.5);
72 hold off
73 set(gca,'Xlim',[-4,4])
74 set(gca,'Xtick',(-4:0.5:4))
75 set(gca,'Ylim',[-4,4])
76 set(gca,'Ytick',(-4:0.5:4))
77 title('Contour Plot of m(\bf p) for \bf x = [0,-1]
        ','interpreter','latex')
78 legend([circ1,p,x0,x1],{'Trust Region','\bf p','x_
        {0}','x_{k+1}'} )
79 colorbar
80 axis equal

```

```

81
82 %cholesky
83 function L = cholesky(~,A)
84 n = length(A);
85 L = zeros(n,n);
86 for i = 1:n
87     L(i,i) = sqrt(A(i,i));
88     if A(i,i) < 0
89         break
90     end
91     for j = i+1:n
92         L(j,i) = A(j,i)/(L(i,i));
93         for k = i+1:j
94             A(j,k) = A(j,k) - L(j,i)*L(k,i);
95         end
96     end
97 end
98 end
99
100 function f = h(x)
101 f = 10*(x(2)-x(1)^2)^2+(1-x(1))^2;
102 end
103
104 function nablaf = grad_f(x)
105 nablaf = [40*x(1)*(x(1)^2-x(2))+2*(x(1)-1), 20*(x(2)-x
106             (1)^2)];
107
108 function A = hess_f(x)
109 A = [120*x(1)^2-40*x(2)+2, -40*x(1);...
110      -40*x(1), 20];
111 end

```