

Optimization Homework 3

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Contents

1	Written description of program	2
2	Did the methods perform as well as I expected?	2
3	Test Results Quadratic function	3
3.1	Results for Quasi-Newton Method	3
3.2	First five steps for Steepest Descent	3
4	Test Results Rosenbrocks function	4
4.1	Results for Quasi-Newton Method	4
4.2	First 100 Steps for Steepest Descent	4
5	Code	5
5.1	fminun	5
5.2	fminunDriver	8
5.3	Rosenbrock Contour plot	10

1 Written description of program

I implemented Steepest Descent and also the Quasi-Newton method with a BFGS update. I used recursion to call the line updates so that I wouldn't have to fiddle around with while or for loops.

The line search was implemented by first taking a step in the search direction as calculated by the chosen algorithm with $\alpha = 0.7$. If this step is too big, the program will start the line search over from the starting point and use $\alpha_{current}/100$. This will keep happening until the first step is smaller than the initial value. After this, the line search keeps searching by doubling alpha each step, always from the same starting point. Once the value of the alpha is bigger than the alpha of the previous value, it finds the value of the point at $\alpha_{current} * 0.75$, halfway between the last point and the second to last point. It then takes the minimum point and the two points around that one, does a quadratic fit, and finds the alpha of the minimum function value of that fit. It then takes a step and finds a new x_0 , and checks the gradient at this point. It repeats this process again until each component of the gradient is under the specified tolerance.

2 Did the methods perform as well as I expected?

For the quadratic equation, the Quasi-Newton algorithm converged with three line searches, which is exactly what we would predict with 3 variables. We know that since Quasi-Newton approximates the algorithm as a quadratic, if the function is a quadratic then it will converge very quickly. Steepest Descent converged in 21 line searches, which is fine. This was not as good as Quasi-Newton, but the function was not elliptical enough to cause the algorithm to take a while to converge.

For the Rosenbrock function, Steepest Descent oscillated back in forth as seen in the figure on page 5. This is predictable because it always tries to go to the minimum but it overshoots it. The Quasi-Newton method had some erratic behavior but converged pretty quickly.

3 Test Results Quadratic function

3.1 Results for Quasi-Newton Method

	Starting Point	Function Value	Search Direction	Step Length	Nobj for search
1	10 10 10	520	-0.99426 0.06414 -0.08552	7.8185	7
2	2.226 10.502 9.3313	154.343	-0.1624 -0.5596 -0.8362	18.3067	9
3	-0.7467 0.2560 -5.977	-14.399	0.05846 0.66920 -0.78849	3.270	6
4	-0.5556 2.4444 -8.5556	-22.3889			1

TOTAL OBJ CALLS = 23

TOTAL GRAD CALLS = 4

3.2 First five steps for Steepest Descent

	Starting Point	Function Value	Search Direction	Step Length	Nobj for search
1	10 10 10	520	-0.99426 0.06414 -0.08552	7.8185	7
2	2.226 10.502 9.3313	154.3	0.0335 -0.5723 -0.8194	12.5261	8
3	2.6466 3.3328 -0.9320	38.88	-0.9765 0.1557 -0.1487	2.5121	6
4	0.1934 3.7239 -1.3057	1.5472	0.1236 -0.1598 -0.9793	4.3808	7
5	0.7352 3.0236 -5.5961	-12.99	-0.9819 0.1229 -0.1441	0.9798	4

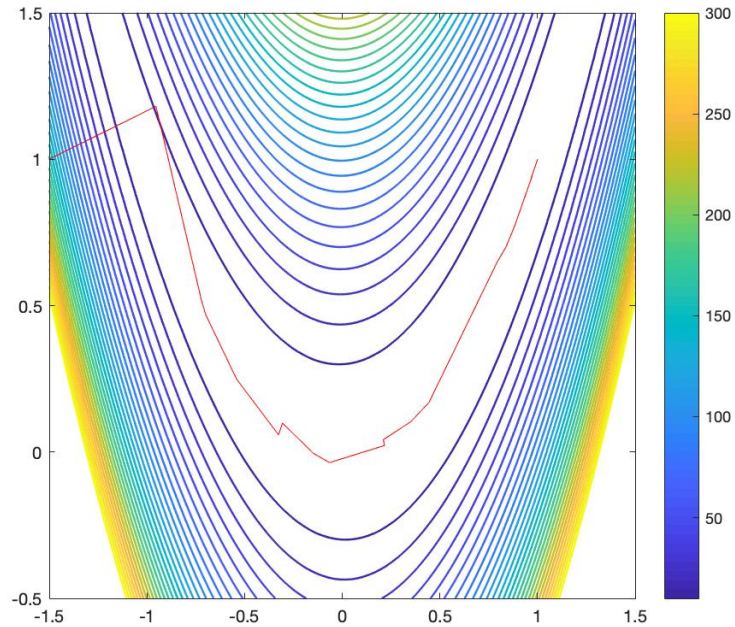
TOTAL OBJ CALLS = 166

TOTAL GRAD CALLS = 24

4 Test Results Rosenbrocks function

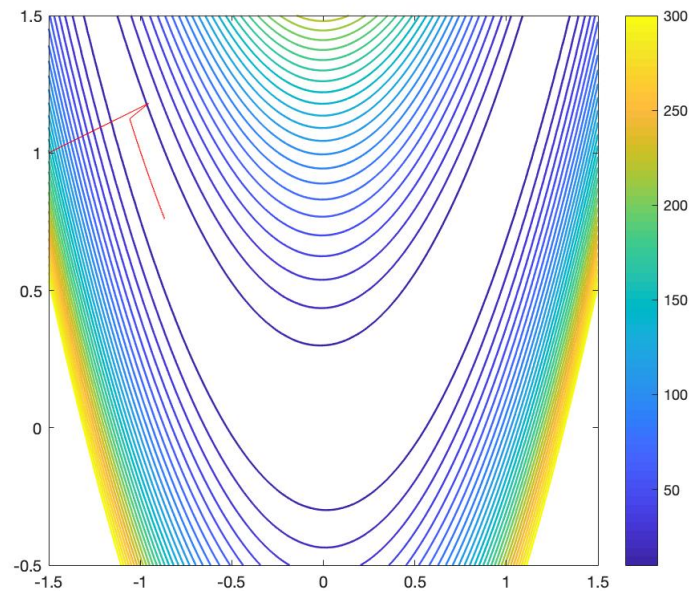
4.1 Results for Quasi-Newton Method

TOTAL OBJ CALLS = 162, TOTAL GRAD CALLS = 22

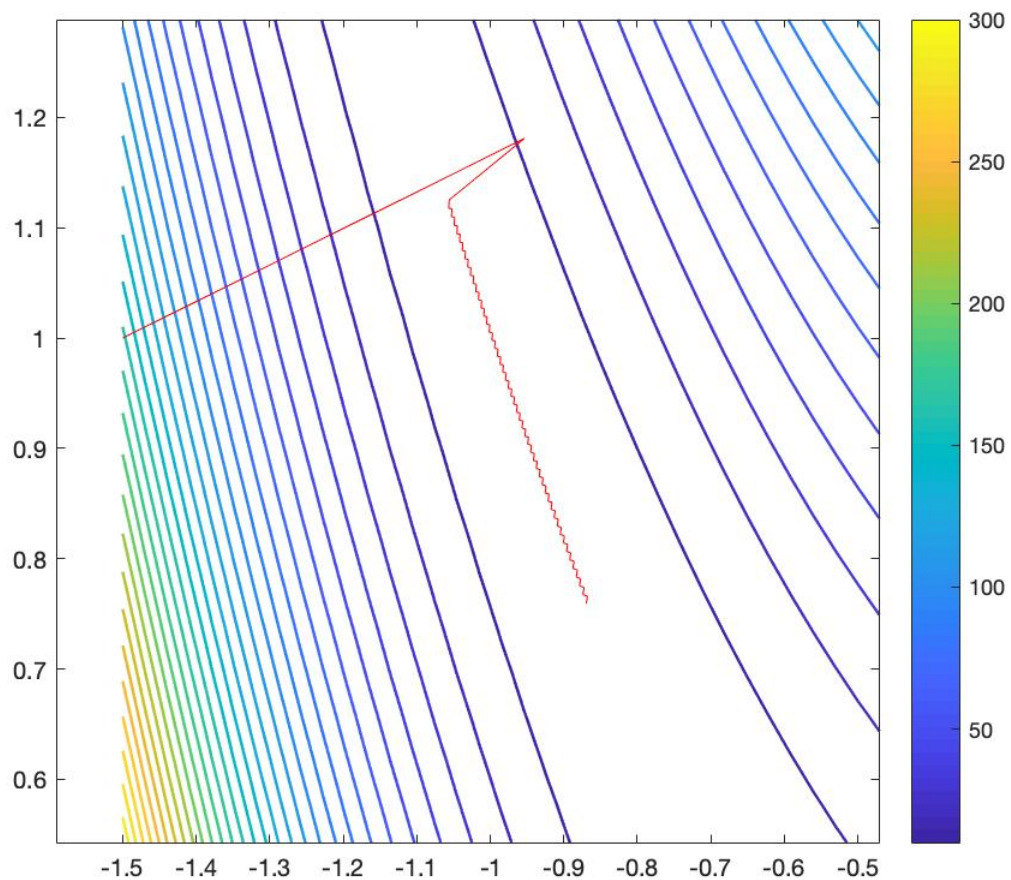


4.2 First 100 Steps for Steepest Descent

TOTAL OBJ CALLS TO CONVERGE = 3931, TOTAL GRAD CALLS TO CONVERGE = 559



Although its hard to see here, the algorithm is taking lots of tiny steps around the line. In the zoomed up view you can see a very small zig zag pattern.



5 Code

5.1 fminun

```

1 function [xopt, fopt, exitflag] = fminun(obj, gradobj, x0, stoptol,
    algoflag)
2     %—— Set constants ——%
3     [n,~] = size(x0);
4     stoptol_vector = stoptol*ones(1,n);
5     alphaInitial = 0.7;
6     firstflag = 1;
7     %—— Set globals ——%
8     global f_a_nobj_history
9     global x_history
10    global s_history
11    %—— Set initial variables ——%
12    allx = [];
13    allf = [];
14    alla = [];
15    alls = [];
16    nobjlocal = [];

```

```

17 N = eye(n);
18 %—— Update Direction ——%
19 [xopt, fopt, exitflag] = updateDirection(obj, gradobj, x0,...
20 stoptol_vector, algoflag, alphaInitial,...
21 N, x0, 0, firstflag, allx, allf, alla, alls, nobjlocal);
22 % Plot Rosenbrock contour plot
23 %     if algoflag == 1
24 %         ContourPlotRosenbrock(x_history(:,1:100))
25 %     else
26 %         % Show all steps
27 %         ContourPlotRosenbrock(x_history)
28 %     end
29 end
30
31 function [xopt,fopt,exitflag] = updateDirection(obj, gradobj, x0,...
32 stoptol_vector, algoflag, alphaInitial, N, xprev, gradprev, firstflag
33 ,...
34 ,allx, allf, alla, alls, nobjlocal)
35 %—— Initialize variables ——%
36 global nobj
37 nobj_at_start = nobj;
38 f = obj(x0);
39 alpha = alphaInitial;
40 grad = gradobj(x0);
41 %—— Check if gradient meets stoptolerance ——%
42 if abs(grad) < stoptol_vector
43     xopt = x0;
44     fopt = f;
45     exitflag = 0;
46     global f_a_nobj_history
47     global x_history
48     global s_history
49     f_a_nobj_history = [allf;alla;nobjlocal];
50     x_history = allx;
51     s_history = alls;
52     return
53 else
54 %—— Check if nobj has been called too many times ——%
55 %     if nobj > 1000
56 %         exitflag = 1;
57 %         xopt = 'algorithm called obj more than 1000 times';
58 %         fopt = 'algorithm called obj more than 1000 times';
59 %         return
60 %     end
61 %—— Update search direction —— %
62 if algoflag == 1
63     s = getSdDirection(grad);
64 elseif algoflag == 2
65     % First iteration of quasi-newton method runs as steepest descent
66     if firstflag == 1
67         s = getSdDirection(grad);
68         firstflag = 0;
69     else
70         [s,N] = getQNdirection(grad,N,grad-gradprev,x0-xprev);
71     end
72 else

```

```

72     error('algorithm not defined')
73 end
74 %—— Do a line search in new direction ——%
75 [a4,fn4] = searchLine(x0,f,s,obj,alpha);
76 [a3,fn3] = takeBestThreePoints(a4,fn4);
77 alpha = quadraticFit(a3,fn3);
78 %—— Record variable history ——%
79 allx = [allx x0];
80 allf = [allf f];
81 alls = [alls s];
82 alla = [alla alpha];
83 nobjlocal = [nobjlocal (nobj-nobj_at_start)];
84 %—— Set new variables for new search direction ——%
85 xnew = x0 + alpha*s;
86 [xopt, fopt, exitflag] = updateDirection(obj, gradobj, xnew,
    stoptol_vector, algoflag, alphaInitial, N,x0,grad,firstflag,allx,
    allf,alla,alls,nobjlocal);
87 end
88 end
89
90 function [a,fn] = searchLine(x0,f,s,obj,alpha)
91     xnew = x0 + alpha*s;
92     fnnew = obj(xnew);
93     if fnnew < f
94         [a,fn] = searchLineHelper(x0,fnnew,s,obj,alpha*2,[0, alpha],[f,fnnew])
95         ;
96     else
97         % If the step was too far, try again with a smaller alpha
98         [a,fn] = searchLine(x0,f,s,obj,alpha/100);
99     return
100 end
101 % Go one step back and return those four points
102 amiddle = a(end)*0.75;
103 fmiddle = obj(x0+amiddle*s);
104 a = [a((end-2):end-1), amiddle, a(end)];
105 fn = [fn((end-2):end-1), fmiddle, fn(end)];
106 end
107
108 function [a,fn] = searchLineHelper(x0,f,s,obj,alpha,ahistory,fhistory)
109     xnew = x0 + alpha*s;
110     fnnew = obj(xnew);
111     if fnnew < f
112         [a,fn] = searchLineHelper(x0,fnnew,s,obj,alpha*2,[ahistory alpha],[fhistory,fnnew]);
113     else
114         a = [ahistory alpha];
115         fn = [fhistory fnnew];
116     end
117 end
118
119 function [astar] = quadraticFit(a,fn)
120     % Calculate Alpha of minimum of the quadratic approximation
121     num = fn(1)*(a(2)^2-a(3)^2)+fn(2)*(a(3)^2-a(1)^2)+fn(3)*(a(1)^2-a(2)^2)
122     ;
123     den = 2*(fn(1)*(a(2)-a(3))+fn(2)*(a(3)-a(1))+fn(3)*(a(1)-a(2)));
124     astar = num/den;

```

```

123 end
124
125 function [s,N] = getQNdirection(grad,N,gamma,delta_x)
126     % BFGS update is used
127     tgamma = transpose(gamma);
128     tdelta_x = transpose(delta_x);
129     first = 1 + (tgamma*N*gamma)/(tdelta_x*gamma);
130     second = (delta_x*tdelta_x)/(tdelta_x*gamma);
131     third = (delta_x*tgamma*N+N*gamma*tdelta_x)/(tdelta_x*gamma);
132     N = N + first*second - third;
133
134     mag = sqrt(grad'*grad);
135     s = -N*(grad/mag);
136 end
137
138 function [a,fn] = takeBestThreePoints(a,fn)
139     [~,min_index] = min(fn);
140     a = a([min_index-1 min_index min_index+1]);
141     fn = fn([min_index-1 min_index min_index+1]);
142 end
143
144 % get steepest descent search direction as a column vector
145 function [s] = getSdDirection(grad)
146     mag = sqrt(grad'*grad);
147     s = -grad/mag;
148 end

```

5.2 fminunDriver

```

1 %—Example Driver program for fminun—%
2 %—Author: Ryan Day—%
3 clear;
4
5 global nobj ngrad
6 nobj = 0; % counter for objective evaluations
7 ngrad = 0.; % counter for gradient evaluations
8
9 algoflag = 2; % 1=steepest descent; 2=BFGS quasi-Newton
10 problem = 2;
11 stoptol = 1.e-3; % stopping tolerance, all gradient elements must be <
    stoptol
12
13 x0_1_starting_points = [[10;10;10] [2;7;9] [100;100;100] [15;9;30]];
14 x0_1 = x0_1_starting_points(:,1); % starting points, set to be column
    vector
15 x0_2_starting_points = [[-1.5; 1] [-3; 1.8]];
16 x0_2 = x0_2_starting_points(:,1);
17
18 if problem == 1
19     % quadratic function
20     x0 = x0_1;
21     obj = @obj1;
22     gradobj = @gradobj1;
23 elseif problem == 2
24     % rosenbrock function
25     x0 = x0_2;

```



```

26         obj = @obj2;
27         gradobj = @gradobj2;
28     end
29     % ----- call fminun -----
30     [xopt, fopt, exitflag] = fminun(obj, gradobj, x0, stoptol, algoflag);
31
32     xopt
33     fopt
34     nobj
35     ngrad
36
37     % Quadratic function to be minimized
38     function [f] = obj1(x)
39         global nobj
40         f = 20+3*x(1)-6*x(2)+8*x(3)+6*x(1)^2-(2*x(1)*x(2))-(x(1)*x(3))+x(2)
41             ^2+0.5*x(3)^2;
42         nobj = nobj +1;
43     end
44     % get gradient as a column vector
45     function [grad] = gradobj1(x)
46         global ngrad
47         %gradient for function 1
48         grad(1,1) = 12*x(1) - 2*x(2) - x(3) + 3;
49         grad(2,1) = 2*x(2) - 2*x(1) - 6;
50         grad(3,1) = x(3) - x(1) + 8;
51         ngrad = ngrad + 1;
52     end
53
54     % Rosenbrock function to be minimized
55     function [f] = obj2(x)
56         global nobj
57         f = 100*(x(2)-x(1)^2)^2+(1-x(1))^2;
58         nobj = nobj +1;
59     end
60
61     % get gradient as a column vector
62     function [grad] = gradobj2(x)
63         global ngrad
64         %gradient for function 1
65         grad(1,1) = 2*x(1) - 400*x(1)*(- x(1)^2 + x(2)) - 2;
66         grad(2,1) = - 200*x(1)^2 + 200*x(2);
67         ngrad = ngrad + 1;
68     end

```

5.3 Rosenbrock Contour plot

```
1 function [] = ContourPlotRosenbrock(coordinates)
2     clf
3
4     %Rosenbrock code from https://www.mathworks.com/matlabcentral/mlc-
5     %downloads/downloads/submissions/23972/versions/22/previews/chebfun/
6     %examples/opt/html/Rosenbrock.html
7     f = @(x,y) (1-x).^2 + 100*(y-x.^2).^2;
8     x = linspace(-1.5,1.5); y = linspace(-1,3);
9     [xx,yy] = meshgrid(x,y); ff = f(xx,yy);
10    levels = 10:10:300;
11    LW = 'linewidth'; FS = 'fontsize'; MS = 'markersize';
12    figure, contour(x,y,ff,levels,LW,1.2), colorbar
13    axis([-1.5 1.5 -0.5 1.5]), axis square, hold on
14    hold on;
15    coordinates
16    plot(coordinates(1,:),coordinates(2,:), 'r');
```