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1 function [x_opt, fval_opt, x_iter, f_iter, alpha] = NewtonsDescent(x0, f, G, H, %
maxiter, grad_tol)
2 % Inputs
3 % x0: initial point
4 % f: function handle
5 % G: function handle for the gradient of f
6 % H: function handle for the hessian of f
7 % maxiter: maximum number of iterations (default 10000)
8 % grad_tol: minimum norm value of gradient (default 1e-4)
9 %
10 % Returns
11 % x_opt: x^*
12 % fval_opt: f(x^*)
13 % x_iter: all iterates k -- column k contains x_k
14 % f_iter: a vector of all function values f(x_k)
15 % alpha: a vector of all step lenghts alpha_k
16
17 % Termination criteria
18 maxiter_default = 10000;
19 grad_tol_default = 1e-4;
20
21 if nargin == 4
22     maxiter = maxiter_default;
23     grad_tol = grad_tol_default;
24 elseif nargin == 5
25     grad_tol = grad_tol_default;
26 end
27
28 nx = size(x0,1); % Number of variables
29
30 % Declare some variables
31 x      = NaN(nx,maxiter);
32 p      = NaN(nx,maxiter);
33 grad   = NaN(nx,maxiter);
34 hessian = NaN(nx,nx,maxiter);
35 alpha  = NaN(1,maxiter);
36 fval   = NaN(1,maxiter);
37
38
39 % Do some calcualtions before the while loop. I.e., do the first iteration:
40 k = 1; % iteration number
41 x(:,k) = x0;
42 % look at the different functions below and finish them before continuing.
43 fval(k) = f(x0);
44 grad(:,k) = G(x0);
45 hessian(:,:,k) = H(x0);
46 p(:,k) = nd(grad(:,k), hessian(:,:,k));
47
48 alpha_0 = 1;
49
50 alpha(k) = linesearch(f, x(:,k), p(:,k), fval(k), grad(:,k), alpha_0);
51 x(:,k+1) = x(:,k) + alpha(k) * p(:,k);
52 grad(:,k+1) = G(x(:,k+1));
53 hessian(:,:,k+1) = H(x(:,k+1));
54 k = k + 1;
55
56 while k < maxiter && norm(grad(:,k),2) >= grad_tol
57     fval(k) = f(x(:,k));    % Evaluate the Rosenbrock function
58     p(:,k) = nd(grad(:,k), hessian(:,:,k));
59     alpha_0 = 1;

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60     alpha(k) = linesearch(f, x(:,k), p(:,k), fval(k), grad(:,k), alpha_0); %✓
Determine alpha using Alg. 3.1
61     x(:,k+1) = x(:,k) + alpha(k) * p(:,k);
62     grad(:,k+1) = G(x(:,k+1));
63     hessian(:,:,k+1) = H(x(:,k+1));
64     k = k + 1;
65 end
66 fval(k) = f(x(:,k)); % Final function value
67
68 % Delete unused space
69 x = x(:,1:k);
70 p = p(:,1:k);
71 grad = grad(:,1:k);
72 hessian = hessian(:,:,1:k);
73 alpha = alpha(1:k);
74 fval = fval(1:k);
75
76 % Return values
77 x_opt = x(:,end);
78 fval_opt = f(x_opt);
79 x_iter = x;
80 f_iter = fval;
81
82 end
83
84 % Function returning the steepest-descent direction based on the gradient
85 % of f
86 function p = nd(grad, hessian)
87     p = -hessian \ grad;
88 end
89
90 % Function implementing Algorithm 3.1, page 37 in N&W
91 function alpha_k = linesearch(f, xk, pk, fk, gradk, alpha_0)
92     alpha = alpha_0;
93     rho = 0.9;
94     c1 = 0.5; % a constant for sufficient decrease
95     while f(xk + alpha*pk) > fk + c1*alpha*(gradk.')*pk
96         alpha = rho * alpha;
97     end
98     alpha_k = alpha;
99 end
100

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