

$$\begin{cases} \min & 2x_1^2 + x_1x_3 + x_2^2 + 2x_2x_3 + 3x_3^2 + x_3x_4 + 2x_4^2 - 5x_1 - 4x_3 + 3x_4 \\ & x \in \mathbb{R}^4 \end{cases}$$

```
clear;clc; close all;
```

a) Do global optimal solutions exist? Why?

```
% Form of the objective function is quadratic function
% Calculate the Hessian. If Hessian is positive definite or positive semidefinite
```

```
H = hessian_matrix();
```

```
f = [-5
      0
     -4
      3];
```

```
eigenvalue = eig(H);
display(eigenvalue)
```

```
eigenvalue = 4x1
    1.0608
    3.5933
    4.0000
    7.3460
```

```
% As result we have a positive definite objective function
% Means Global Optimum exist and is unique
```

b) Is it a convex problem? Why?

```
% It is strictly convex
```

```
x = quadprog(H,f);
```

Minimum found that satisfies the constraints.

Optimization completed because the objective function is non-decreasing in feasible directions, to within the value of the optimality tolerance, and constraints are satisfied to within the value of the constraint tolerance.

<stopping criteria details>

```
display(x);
```

```
x = 4x1
    1.0000
   -1.0000
    1.0000
```

-1.0000

```
x0 = [0
      0
      0
      0];
alpha = 0.1;
gamma = 0.9;
t_bar = 1;
tolerance = 1e-6;

% Gradient Method with line search
x_gradient_method = gradient_method(x0, tolerance,H);
```

Gradient method with exact line search

Iteration	f(x)	grad f(x)
0	0.00000000000000	7.0711e+00
1	-5.0403225806452	2.4686e+00
2	-5.5976694221960	1.1368e+00
3	-5.7883000924345	9.8085e-01
4	-5.8850831814683	6.0051e-01
5	-5.9375039953765	5.3213e-01
6	-5.9660079447169	3.2658e-01
7	-5.9815113436573	2.8943e-01
8	-5.9899438100569	1.7763e-01
9	-5.9945303240821	1.5742e-01
10	-5.9970249811449	9.6616e-02
11	-5.9983818534549	8.5624e-02
12	-5.9991198717155	5.2550e-02
13	-5.9995212882297	4.6572e-02
14	-5.9997396232310	2.8583e-02
15	-5.9998583781180	2.5331e-02
16	-5.9999229702499	1.5546e-02
17	-5.9999581026440	1.3778e-02
18	-5.9999772115522	8.4559e-03
19	-5.9999876051044	7.4939e-03
20	-5.9999932582755	4.5993e-03
21	-5.9999963330995	4.0760e-03
22	-5.9999980055312	2.5016e-03
23	-5.9999989151857	2.2170e-03
24	-5.9999994099572	1.3606e-03
25	-5.9999996790690	1.2058e-03
26	-5.9999998254420	7.4007e-04
27	-5.9999999050559	6.5587e-04
28	-5.9999999483588	4.0253e-04
29	-5.9999999719118	3.5674e-04
30	-5.9999999847225	2.1894e-04
31	-5.9999999916904	1.9403e-04
32	-5.9999999954803	1.1908e-04
33	-5.9999999975417	1.0554e-04
34	-5.9999999986629	6.4772e-05
35	-5.9999999992727	5.7403e-05
36	-5.9999999996044	3.5230e-05
37	-5.9999999997848	3.1222e-05
38	-5.9999999998830	1.9162e-05
39	-5.9999999999363	1.6982e-05

40	-5.9999999999654	1.0422e-05
41	-5.9999999999812	9.2367e-06
42	-5.9999999999898	5.6689e-06
43	-5.9999999999944	5.0239e-06
44	-5.9999999999970	3.0834e-06
45	-5.9999999999984	2.7326e-06
46	-5.9999999999991	1.6771e-06
47	-5.9999999999995	1.4863e-06
48	-5.9999999999997	9.1218e-07

```
% Conjugate Gradient Method
```

```
x_conjugate_gradient = conjugate_method(x0, tolerance, H);
```

Conjugate Gradient method

Iteration	f(x)	grad f(x)
0	0.00000000000000	7.0711e+00
1	-5.0403225806452	2.4686e+00
2	-5.6669616780246	9.2779e-01
3	-5.9998774890506	2.9667e-02
4	-6.00000000000000	2.8436e-15

```
% Newton Method with Armijo line search
```

```
x_newton_method = newton_method(x0, alpha, gamma, t_bar, tolerance);
```

Newton Method

Iterations	f(x)	grad f(x)
0	0.00000000000000	7.0711e+00
1	-6.00000000000000	0.0000e+00

c) Find the global minimum by using the gradient method with exact line search starting from the point (0,0,0,0) [Use $\|\nabla f(x)\| < 10^{-6}$ as stopping criterion]. How many iterations are needed?

```
function alpha = line_search(g, d, H)
```

```
    alpha = (-g'*d)/(d'*H*d);
```

```
end
```

```
function x_opt = gradient_method(x0, tolerance, H)
```

```
    fprintf('Gradient method with exact line search\n\n');
```

```
    fprintf('Iteration \t f(x) \t\t ||grad f(x)||\n\n');
```

```
    k = 0 ;
```

```
    % starting point
```

```
    x = x0;
```

```
    while true
```

```
        [f, g, H] = f_(x);
```

```
        fprintf('%2.0f \t %2.13f \t %1.4e\n',k,f,norm(g));
```

```

    % stopping criterion
    if norm(g) < tolerance
        break
    end

    % search direction
    d = -g;

    % step size
    alpha = line_search(g, d, H);

    % new point
    x = x + alpha*d ;
    k = k + 1 ;
end
x_opt = x;
end

```

- d) Find the global minimum by using the conjugate gradient method starting from the point $(0,0,0,0)$. How many iterations are needed? Write the point found by the method at any iteration.

```

function beta = conjugate_function(g,d,H)
    beta = (g'*H*d)/(d'*H*d);
end
% x0: Initial point
% d: search direction
function x_opt = conjugate_method(x0, tolerance, H)
    fprintf('Conjugate Gradient method \n\n');
    fprintf('Iteration \t f(x) \t\t ||grad f(x)||\n\n');

    x = x0;
    k = 0;
    g = gradient_function(x);
    while true
        f = objective_function(x);
        fprintf('%2.0f \t %2.13f \t %1.4e\n',k,f,norm(g));

        % Stop Criteria
        if norm(g) < tolerance
            break;
        end

        if k == 0
            d = -g;
        else
            % conjugate function
            beta = conjugate_function(g,d,H);
            d = -g + beta * d;
        end
    end
    x_opt = x;
end

```

```

end

% step size
alpha = line_search(g,d,H);

% new point
x = x + alpha .* d;

g = gradient_function(x);
k = k+1;

end
x_opt = x;
end

```

e) Find the global minimum by using the Newton method starting from the point (0,0,0,0). How many iterations are needed?

```

function x_opt = newton_method(x0,alpha,gamma,t_bar, tolerance)
fprintf("Newton Method\n\n");
fprintf("Iterations \t f(x) \t\t ||grad f(x)||\n");
k = 0;
x = x0;
while true
    [f, g, H] = f_(x);

    fprintf("%2.0f \t %2.13f \t %1.4e\n", k,f,norm(g));

    % Stop Criteria
    if norm(g) < tolerance
        break;
    end

    % Search direction  $H*d = -g$ 
    % Example  $Ax=b$  Linear equation  $x=A\b$ 
    d = -H\g;

    % Step size with Armijo line search
    t = t_bar;

    while f_(x+t*d)> f + alpha*t*d'*g
        t = gamma*t;
    end
    x = x + t*d;
    k = k+1;
end
x_opt = x;
end

```

```

function [f,g,H] = f_(x)
    f = objective_function(x);
    g = gradient_function(x);
    H = hessian_matrix();
end

function H = hessian_matrix()
    H = [4 0 1 0
          0 2 2 0
          1 2 6 1
          0 0 1 4];
end

function f = objective_function(x)
% x: Input vector
% f: output scalar value of the objective function
    x1 = x(1);
    x2 = x(2);
    x3 = x(3);
    x4 = x(4);

    f = 2*x1^2+x1*x3+x2^2+2*x2*x3+3*x3^2+x3*x4+2*x4^2-5*x1-4*x3+3*x4;
end

function g = gradient_function(x)
    x1 = x(1);
    x2 = x(2);
    x3 = x(3);
    x4 = x(4);

    g = [4*x1+x3-5
          2*x2+2*x3
          x1+2*x2+6*x3+x4-4
          x3+4*x4+3];
end

```