Q1 - Assignment 9 | ME7223

Saarthak Marathe - ME17B162

In [126]:

import numpy as np
import matplotlib.pyplot as plt
from sympy import symbols, solve
from scipy.optimize import minimize
import scipy
import math
import matplotlib.pyplot as plt

In [127]:

```
def f(x):
 2
        return (x[0]-1)**2+(x[1]-2)**2 - 4
 3
 4
    def g1(x):
 5
        return x[0]+2*x[1]-5
 6
 7
    def g2(x):
 8
        return 4*x[0]+3*x[1]-10
 9
10
    def g3(x):
11
        return 6*x[0]+x[1]-7
12
    def diff_f(x):
13
14
        df0 = 2*(x[0]-1)
        df1 = 2*(x[1]-2)
15
16
        return np.array([df0,df1])
17
    def g_valcheck(x):
18
19
        check = [0]*5
        if g1(x) < 0 and g2(x) < 0 and g3(x) < 0 and x[0] > 0 and x[1] > 0:
20
21
            return check;
22
        elif g1(x) == 0:
23
            check[0] = 1
24
        elif g2(x) == 0:
25
            check[1] = 1
26
        elif g3(x) == 0:
27
            check[2] = 1
        elif x[0] == 0:
28
29
            check[3] = 1
30
        elif x[1] == 0:
31
            check[4] = 1
32
        else:
33
            return False;
34
        return check;
35
36
    def diff_g1(x):
37
        dg1 \times 0 = 1
38
        dg1 x1 = 2
39
        return np.array([dg1_x0,dg1_x1])
40
    def diff g2(x):
41
42
        dg2_x0 = 4
43
        dg2 x1= 3
44
        return np.array([dg2_x0,dg2_x1])
45
46
    def diff_g3(x):
47
        dg3 x0 = 6
48
        dg3 x1=1
49
        return np.array([dg3_x0,dg3_x1])
50
51
    def norm(x):
52
        return math.sqrt(x[0]**2 + x[1]**2)
```

Zoutendijk Method

In [128]:

```
def bestdir finder(x, cv):
                         # this keeps the alpha, s1,s2 as variables until solved
   2
                         s = np.array([symbols("s1"), symbols("s2")])
    3
   4
                         a = symbols("a")
    5
                         grad_fun = np.dot(s, diff_f(x)) + a
    6
   7
                         g1_grad, g2_grad, g3_grad = diff_g1(x), diff_g2(x), diff_g3(x)
   8
   9
                         x1_grad = np.array([-1, 0])
10
                         x2_grad = np.array([0, -1])
11
                         constraint_fun = cv[0]*np.dot(s, g1_grad) + cv[1]*np.dot(s, g2_grad) + cv[2]*np.dot(s, g2_grad) + cv
12
13
                         magnitude_fun = np.dot(s.T, s) - 1
14
15
                         s_fun = lambda check: float(-check[2] + max(0, grad_fun.evalf(subs = {s[0]: check[4]})
16
17
                         res = minimize(s_fun, [1, 1, 1])
18
                         best_s = res['x'][:2]
19
20
                         return (best_s)
21
22
           def best_step(x, s):
                         1 = symbols("1")
23
24
                         x_symb = x + 1 * s
                         f_{uni} = lambda 1: f(x + 1 * s)
25
26
                         residual = scipy.optimize.minimize(f_uni, 0)
27
                         opt = residual['x']
28
                         x_new = x + opt * s
29
                         while g_valcheck(x_new) == False:
                                      opt = opt * 0.8
30
31
                                      x_new = x + opt*s
32
                         return opt
```

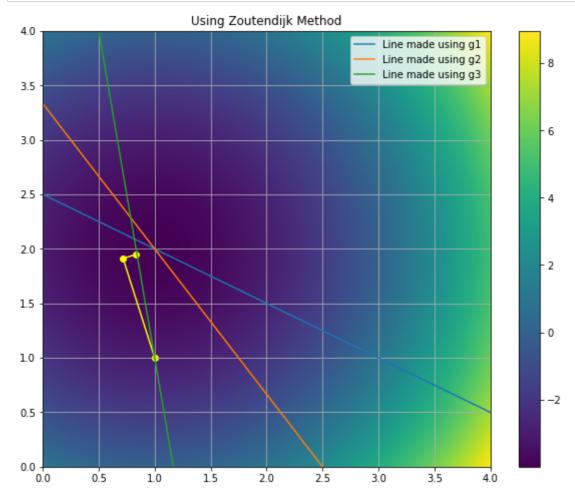
In [129]:

```
1 \mid x = np.array([1,1])
 2 coords = [x.tolist()]
 3 x = list(x)
 4 print(0, '- Coordinate:', x, '- Objective Function:', f(x))
 5 | i = 0
   max_iter = 2
   eps = [1e-6]*3
 7
   err1, err2 = float('inf'), float('inf')
10
   while err1 > eps[1] and err2 > eps[2] and i<2:
11
        vec_check = g_valcheck(x)
        if np.sum(vec_check) == 0:
12
13
            s = - diff_f(x)
            len_step = best_step(x, s)
14
            x_new = x + len_step * s
15
16
            error2 = abs((f(x) - f(x_new)) / f(x))
            error3 = norm(x - x_new)
17
18
            x = x_new
19
            coords.append(x)
20
21
       else:
22
            s = bestdir_finder(x, vec_check)
23
            len_step = best_step(x, s)
24
            x_new = x + len_step * s
25
            err1 = abs((f(x) - f(x_new)) / f(x))
26
            err2 = norm(x - x_new)
27
            x = x_new
28
            coords.append(x)
29
        i = i + 1
        print(i, '- Coordinate:', x, '- Objective Function:', f(x))
30
```

```
0 - Coordinate: [1, 1] - Objective Function: -3
1 - Coordinate: [0.71745035 1.91251145] - Objective Function: -3.91251144711
33737
2 - Coordinate: [0.83318267 1.94834676] - Objective Function: -3.96950392254
4454
```

In [130]:

```
coords = np.array(coords)
 2
   xmin = 0
 3
   xmax = 4
4
   ymin = 0
   ymax = 4
 5
   x = np.linspace(xmin, xmax, 600)
   y = np.linspace(ymin,ymax,600)
 7
   xx, yy = np.meshgrid(x, y)
   zz = np.array([f(xy) for xy in np.c_[xx.ravel(), yy.ravel()]]).reshape(xx.shape)
   plt.figure(figsize=(10,8))
11 plt.pcolormesh(xx, yy, zz)
   plt.colorbar()
12
   plt.plot(coords[:,0], coords[:,1], color = 'yellow')
13
   plt.scatter(coords[:,0], coords[:,1], color = 'yellow')
15
16 y1 = (5-x)/2
17 y2 = (10-4*x)/3
18 \ y3 = 7 - 6*x
   plt.plot(x,y1, label = 'Line made using g1')
19
   plt.plot(x,y2, label = 'Line made using g2')
20
   plt.plot(x,y3, label = 'Line made using g3')
22 plt.xlim(xmin, xmax)
23
   plt.ylim(ymin, ymax)
24 plt.grid()
25 plt.title('Using Zoutendijk Method')
26 plt.legend()
27 plt.show()
```



Rosen Gradient Projection Method

In [131]:

```
1
   def matrix n(vec):
 2
        #gradients of g1,g2,g3,x0,x1 combined
 3
        diff_mat = np.array([[1,4,6,-1,0],[2,3,1,0,-1]])
 4
        1 = np.sum(vec)
 5
        N = np.zeros((2,1))
 6
        j = 0
 7
        for i in range(len(vec)):
 8
            if vec[i] == 1:
 9
                N[:,j] = diff_mat[:,i]
                j += 1
10
11
        return N;
12
   def bestdir_finder(x, vec):
13
        N = matrix n(vec)
14
        P = np.eye(2) - np.dot(np.linalg.inv(np.dot(N.T, N)),N.T)
15
        s = -(np.dot(P.T,diff_f(x)))
16
        return s
17
```

In [132]:

```
1 \mid x = np.array([1,1])
 2 coords = [x.tolist()]
 3 x = list(x)
 4 print(0, '- Coordinate:', x, '- Objective Function:', f(x))
   max_iter = 1
   eps = 1e-6
 7
   err = norm(diff_f(x))
10
   while err > eps and i<max iter:
        vec_check = g_valcheck(x)
11
        if np.sum(vec_check) == 0:
12
            s = - diff_f(x)
13
            len_step = best_step(x, s)
14
            x_new = x + len_step * s
15
            err = norm(diff_f(x_new))
16
            x = x_new
17
            coords.append(x)
18
19
20
        else:
            s = bestdir_finder(x, vec_check)
21
22
            step_length = best_step(x, s)
            x \text{ new} = x + 1\text{en step} * s
23
24
            error = norm(diff_f(x_new))
25
            x = x_new
26
            coords.append(x)
27
        i += 1
28
29
        print(i, '- Coordinate:', x, '- Objective Function:', f(x))
```

```
0 - Coordinate: [1, 1] - Objective Function: -3
1 - Coordinate: [0.93357838 1.3985297 ] - Objective Function: -3.63382164211
9582
```

In [133]:

```
coords = np.array(coords)
 2
   xmin = 0
 3
   xmax = 4
4
   ymin = 0
   ymax = 4
 5
   x = np.linspace(xmin, xmax, 600)
   y = np.linspace(ymin,ymax,600)
 7
8 \times x, yy = np.meshgrid(x, y)
   zz = np.array([f(xy) for xy in np.c_[xx.ravel(), yy.ravel()]]).reshape(xx.shape)
   plt.figure(figsize=(10,8))
11 plt.pcolormesh(xx, yy, zz)
   plt.colorbar()
12
   plt.plot(coords[:,0], coords[:,1], color = 'yellow')
13
   plt.scatter(coords[:,0], coords[:,1], color = 'yellow')
15
16 y1 = (5-x)/2
17 y2 = (10-4*x)/3
18 \ y3 = 7 - 6*x
   plt.plot(x,y1, label = 'Line made using g1')
19
   plt.plot(x,y2, label = 'Line made using g2')
20
   plt.plot(x,y3, label = 'Line made using g3')
22 plt.xlim(xmin, xmax)
23
   plt.ylim(ymin, ymax)
24 plt.title('Using Rosen Gradient Method')
25 plt.grid()
26 plt.legend()
27 plt.show()
```

