

# Homework 10

Zhen

December 6, 2021

## Problem 1

The KKT conditions are:

$$2x_1 + 2\mu_1(x_1 - 1) + 2\mu_2(x_1 - 1) = 0$$

$$2x_2 + 2\mu_1(x_2 - 1) + 2\mu_2x_2 = 0$$

$$\mu_i \geq 0, \quad i = 1, 2$$

$$\mu_i g_i(\mathbf{x}) = 0, \quad i = 1, 2$$

**Case I:**  $g_2$  is inactive and  $g_1$  is active. Then  $\mu_2 = 0$  and  $g_1(\mathbf{x}) = 0$ .

$$\begin{cases} x_1 = x_2 = 1 - \frac{\sqrt{2}}{2} \\ \mu_1 = \sqrt{2} - 1 \end{cases} \quad \text{or} \quad \begin{cases} x_1 = x_2 = 1 + \frac{\sqrt{2}}{2} \\ \mu_1 = -1 - \sqrt{2} \end{cases} \quad (\text{contradiction to } \mu_1 \geq 0)$$

By the sufficiency of KKT conditions for convex problems,  $\mathbf{x}^* = (1 - \frac{\sqrt{2}}{2}, 1 - \frac{\sqrt{2}}{2})$ ,  $\mu_1 = \sqrt{2} - 1$ ,  $\mu_2 = 0$ .

**Note that:** this theorem (the sufficiency of KKT conditions for convex problems) will be used for many times. For the sake of convenience we denote it as “Thm 1”.

## Problem 2

(a) The feasible set is  $\{(x_1, x_2) : x_1 = 1 \wedge x_2 = 0\}$ . Then  $\mathbf{x}^* = (1, 0)$ ,  $f^* = 1$ .

(b) The KKT conditions are:

$$2x_1 + 2\mu_1(x_1 - 1) + 2\mu_2(x_1 - 1) = 0$$

$$2x_2 + 2\mu_1(x_2 - 1) + 2\mu_2(x_2 + 1) = 0$$

$$\mu_i \geq 0, \quad i = 1, 2$$

$$\mu_1[(x_1 - 1)^2 + (x_2 - 1)^2 - 1] = 0$$

$$\mu_2[(x_1 - 1)^2 + (x_2 + 1)^2 - 1] = 0$$

$x_1 = 1$  and  $x_2 = 0$  will lead to contradiction. Thus there does not exist Lagrange multipliers satisfying the KKT conditions and both  $g$  are active.

$$\nabla g(\mathbf{x}^*) = \begin{bmatrix} 0 & 1 \\ 0 & 1 \end{bmatrix}$$

Thus  $\mathbf{x}^*$  is not regular.

## Problem 3

Let

$$\mathbf{g}(\mathbf{x}^*) = \begin{bmatrix} g_1(\mathbf{x}^*) \\ g_2(\mathbf{x}^*) \\ g_3(\mathbf{x}^*) \\ g_4(\mathbf{x}^*) \end{bmatrix} = \begin{bmatrix} x_1^2 - x_2 \\ x_1 + x_2 - 6 \\ -x_1 \\ -x_2 \end{bmatrix}$$

The KKT conditions are:

$$2(x_1 - \frac{9}{4}) + 2\mu_1 x_1 + \mu_2 - \mu_3 = 0$$

$$2(x_2 - 2) - \mu_1 + \mu_2 - \mu_4 = 0$$

$$\mu_i \geq 0, \quad i = 1, 2, 3, 4$$

$$\mu_i g_i(\mathbf{x}) = 0, \quad i = 1, 2, 3, 4$$

**Case I:**  $g_1$  is active and  $g_2, g_3, g_4$  are inactive. Then  $x_1^2 = x_2$  and  $\mu_2 = \mu_3 = \mu_4 = 0$ .

Then we have

$$x_1 = \frac{3}{2}, \quad x_2 = \frac{9}{4}, \quad \mu_1 = \frac{1}{2}$$

By Thm 1 and the fact that  $f, \mathbf{g}$  are convex:

$$\mathbf{x}^* = \begin{bmatrix} \frac{3}{2} & \frac{9}{4} \end{bmatrix}, \quad \mu_1 = \frac{1}{2}, \quad \mu_2 = \mu_3 = \mu_4 = 0$$

## Problem 4

(a) The KKT conditions are:

$$\mathbf{x} - \mathbf{z} + \lambda \mathbf{y} - \boldsymbol{\mu} = \mathbf{0}$$

$$\mathbf{y}^T \mathbf{x} = 0$$

$$\boldsymbol{\mu} \geq \mathbf{0}$$

$$\mathbf{M}\mathbf{x} = \mathbf{0}$$

where  $\lambda \in \mathbb{R}$ ,  $\boldsymbol{\mu} = (\mu_1, \dots, \mu_n)^T \in \mathbb{R}$ ,  $\mathbf{M} = \text{diag}(\mu_1, \dots, \mu_n)$ .

Then we have:

- If  $x_i^* = 0$  and  $\mu_i \geq 0$ , then  $x_i = \mu_i + z_i - \lambda y_i = 0$  which means that  $z_i - \lambda y_i \leq 0$ . Thus  $x_i = (z_i - \lambda y_i)^+$ .
- If  $\mu_i = 0$ , then  $x_i = z_i - \lambda y_i > 0$ .

Thus  $x_i = (z_i - \lambda y_i)^+$ .

And by the fact that  $\mathbf{y}^T \mathbf{x} = 0$ ,  $\sum_{i=1}^n y_i (z_i - \lambda y_i)^+ = 0$ .

(b) My code is in the next page (Also in the file “code.py”).

The final result is:  $\mathbf{x}^* = (\frac{1}{3}, \frac{4}{3}, \frac{5}{3})^T$  and  $\lambda = \frac{2}{3}$ .

```

1  import numpy as np
2
3  def solve(z,y):
4      """
5          compute lamda for a particular optimization problem:
6          \min 0.5*\norm{x-z}^2
7          s.t. y^Tx = 0
8              x >= 0
9
10         Inputs:
11         - z: vector in objective function, of shape (N,) or (N,1)
12         - y: vector in equality constraint function, of any given shape (N,) or ...
13               (N,1)
14
15         Returns a tuple of:
16         - x: the optimal point of the problem, of shape (N,) or (N,1). x = ...
17               (z-lamda*y)^+.
18         - lamda: the Lagrange multiplier of the equality constraint function in ...
19               KKT conditions.
20
21         """
22
23         shape = z.shape
24         z = z.reshape(-1)
25         y = y.reshape(-1)
26         # initialize some parameters
27         p = np.sum(y>0) # number of the elements > 0 in y
28         m = len(y) - p
29         u = np.sort(z[y==1]) # u = {z_i: i \in I_+}
30         w = np.sort(-z[y==1])
31         u = np.append([-np.inf],u)
32         u = np.append(u,[np.inf])
33         w = np.append([-np.inf],w)
34         w = np.append(w,[np.inf])
35
36         U = np.zeros(p+1)
37         W = np.zeros(m+1)
38         for i in range(p):
39             U[p-i-1] = U[p-i] + u[p-i]
40         for i in range(m):
41             W[i+1] = W[i] + w[i+1]
42         # initialization of lamda, l and k
43         lam = -1
44         k = 0
45         l = 0
46         # Nlog(N) algortithm
47         while (k<=p) and (l<=m):
48             lam = (U[k] + W[l]) / (p - k + l)
49             if (u[k] <= lam <= u[k+1]) and (w[l] <= lam <= w[l+1]):
50                 break
51             if u[k+1] < w[l+1]:
52                 k = k+1
53             else:
54                 l = l+1
55         return (((z - lam * y) > 0) * (z - lam * y)).reshape(shape), lam
56
57 z_p4 = np.array([1.0, 2.0, 1.0])
58 y_p4 = np.array([1.0, 1.0, -1.0])
59 x,lamda = solve(z_p4, y_p4)
60 # print(x,lamda)

```