Question 1

First we construct the Lagrangian as below:

The KKT conditions are

$$\begin{cases} \partial_{x_1} \mathcal{L} = 2x_1 + \mu_1 + 2\mu_2 (x_1 - 1) = 0 \\ \partial_{x_2} \mathcal{L} = 2 (x_2 - 1) - \mu_1 + 2\mu_2 x_2 = 0 \\ g_1(\boldsymbol{x}) = x_1 - x_2 - 1 \le 0 \\ g_2(\boldsymbol{x}) = (x_1 - 1)^2 + x_2^2 - 1 \le 0 \\ \mu_1 \ge 0 \\ \mu_2 \ge 0 \\ \mu_1 g_1(\boldsymbol{x}) = \mu_1 (x_1 - x_2 - 1) = 0 \\ \mu_2 g_2(\boldsymbol{x}) = \mu_2 \left[(x_1 - 1)^2 + x_2^2 - 1 \right] = 0 \end{cases}$$

$$(2)$$

For j = 1, 2, we have the following four cases:

Case 1: $g_1(\boldsymbol{x}), g_2(\boldsymbol{x})$ are both inactive.

Then we have $\mu_1 = \mu_2 = 0, \; x_1 = 0, x_2 = 1$

However, $\,x_1=0,x_2=1$ doesn't satisfy $g_2(oldsymbol{x})=(x_1-1)^2+x_2^2-1\leq 0$

Thus we should exclude this case.

Case 2: $g_1(\boldsymbol{x}), g_2(\boldsymbol{x})$ are both active.

Then we have:

$$\begin{cases} g_1(\boldsymbol{x}) = x_1 - x_2 - 1 = 0 \\ g_2(\boldsymbol{x}) = (x_1 - 1)^2 + x_2^2 - 1 = 0 \end{cases}$$
(3)

And induce that $x_1=1\pm rac{\sqrt{2}}{2}, x_2=\pm rac{\sqrt{2}}{2}$, together with the equations below:

$$egin{cases} \partial_{x_1}\mathcal{L} = 2x_1 + \mu_1 + 2\mu_2 \left(x_1 - 1
ight) = 0 \ \partial_{x_2}\mathcal{L} = 2\left(x_2 - 1
ight) - \mu_1 + 2\mu_2 x_2 = 0 \end{cases}$$

we have $\mu_2=-1$, which contradicts the fact that μ_2 should be larger than zero, thus this case should be excluded.

Case 3: $g_1(\boldsymbol{x})$ is active, while $g_2(\boldsymbol{x})$ is inactive.

Then we have:

$$\begin{cases}
g_1(\mathbf{x}) = x_1 - x_2 - 1 = 0 \\
\mu_2 = 0
\end{cases}$$
(5)

With

$$\begin{cases} \partial_{x_1} \mathcal{L} = 2x_1 + \mu_1 + 2\mu_2 (x_1 - 1) = 0 \\ \partial_{x_2} \mathcal{L} = 2(x_2 - 1) - \mu_1 + 2\mu_2 x_2 = 0 \end{cases}$$
 (6)

We have $x_2 = 0$, $\mu_1 = -2$, which contradicts the fact that μ_1 should be larger than zero, thus this case should be excluded.

Case 4: $g_2(\boldsymbol{x})$ is active, while $g_1(\boldsymbol{x})$ is inactive.

Then we have:

$$\begin{cases}
g_2(\mathbf{x}) = (x_1 - 1)^2 + x_2^2 - 1 = 0 \\
\mu_1 = 0
\end{cases}$$
(7)

With

$$\begin{cases} \partial_{x_1} \mathcal{L} = 2x_1 + \mu_1 + 2\mu_2 (x_1 - 1) = 0 \\ \partial_{x_2} \mathcal{L} = 2(x_2 - 1) - \mu_1 + 2\mu_2 x_2 = 0 \end{cases}$$
(8)

We have

$$\begin{pmatrix}
x_1 = 1 - \frac{1}{\sqrt{2}} \\
x_2 = \frac{1}{\sqrt{2}} \\
\mu_1 = 0 \\
\mu_2 = \sqrt{2} - 1
\end{pmatrix}
\begin{pmatrix}
x_1 = 1 + \frac{1}{\sqrt{2}} \\
x_2 = -\frac{1}{\sqrt{2}} \\
\mu_1 = 0 \\
\mu_2 = -\sqrt{2} - 1
\end{pmatrix}$$
(9)

In (2), we have $g_1(oldsymbol{x})=x_1-x_2-1\geq 0$, thus get excluded.

In (1), we have $g_1(oldsymbol{x})=x_1-x_2-1\leq 0$, satisfy the KKT conditions.

Thus the optimal point \boldsymbol{x}^* is $\left(1-\frac{1}{\sqrt{2}},\frac{1}{\sqrt{2}}\right)$ and the Lagrange multipliers are $\mu_1=0,\mu_2=\sqrt{2}-1.$

Question 2

We know the fact that KKT conditions for convex problems are both sufficient and necessary.

First, we transform the problem into the following convex form:

$$egin{align} \min & x_1^2 + x_2^2 \ ext{s.t.} & g_1(oldsymbol{x}) = (x_1 - 1)^2 + (x_2 - 1)^2 + 1 \leq 0 \ & g_2(oldsymbol{x}) = (x_1 - 1)^2 + x_2^2 + 1 \leq 0 \ \end{pmatrix} \ (10)$$

We construct the Lagrangian as below:

The KKT conditions are:

$$\begin{cases} \partial_{x_{1}}\mathcal{L} = 2x_{1} + 2\mu_{1} \left(x_{1} - 1\right) + 2\mu_{2} \left(x_{1} - 1\right) = 0 \\ \partial_{x_{2}}\mathcal{L} = 2x_{2} + 2\mu_{1} \left(x_{2} - 1\right) + 2\mu_{2}x_{2} = 0 \\ g_{1}(\boldsymbol{x}) = \left(x_{1} - 1\right)^{2} + \left(x_{2} - 1\right)^{2} - 1 \leq 0 \\ g_{2}(\boldsymbol{x}) = \left(x_{1} - 1\right)^{2} + x_{2}^{2} - 1 \leq 0 \\ \mu_{1} \left[\left(x_{1} - 1\right)^{2} + \left(x_{2} - 1\right)^{2} - 1 \right] = 0 \\ \mu_{2} \left[\left(x_{1} - 1\right)^{2} + x_{2}^{2} - 1 \right] = 0 \\ \mu_{1} \geq 0 \\ \mu_{2} \geq 0 \end{cases}$$

$$(12)$$

Then we check the following points:

$$(1) (x_1, x_2) = (1, 1)$$

Since
$$g_1(\boldsymbol{x}) = (x_1 - 1)^2 + (x_2 - 1)^2 - 1 < 0$$

$$\therefore \mu_1 = 0$$

Then
$$\partial_{x_1}\mathcal{L}=2x_1+2\mu_1\left(x_1-1
ight)+2\mu_2\left(x_1-1
ight)=0$$
 not holds for $(x_1,x_2,\mu_1)=(1,1,0)$

Thus this point is not an optimal point.

$$(2) (x_1, x_2) = (0, 1)$$

This is not an optimal point since $g_2(oldsymbol{x}) = (x_1-1)^2 + x_2^2 - 1 = 1 > 0$

(3)
$$(x_1,x_2)=(1-rac{\sqrt{2}}{2},1-rac{\sqrt{2}}{2})$$

Since
$$g_2({m x}) = (x_1-1)^2 + x_2^2 - 1 = 1 - \sqrt{2} < 0$$

$$\therefore \mu_2 = 0$$

By solving the KKT conditions, we derive that $\mu_1=\sqrt{2}-1>0$

And it satisfies:

$$\begin{cases} \partial_{x_{1}}\mathcal{L} = 2x_{1} + 2\mu_{1} (x_{1} - 1) + 2\mu_{2} (x_{1} - 1) = 0 \\ \partial_{x_{2}}\mathcal{L} = 2x_{2} + 2\mu_{1} (x_{2} - 1) + 2\mu_{2}x_{2} = 0 \\ g_{1}(\boldsymbol{x}) = (x_{1} - 1)^{2} + (x_{2} - 1)^{2} - 1 \leq 0 \\ \mu_{1} \left[(x_{1} - 1)^{2} + (x_{2} - 1)^{2} - 1 \right] = 0 \\ \mu_{2} \left[(x_{1} - 1)^{2} + x_{2}^{2} - 1 \right] = 0 \end{cases}$$

$$(13)$$

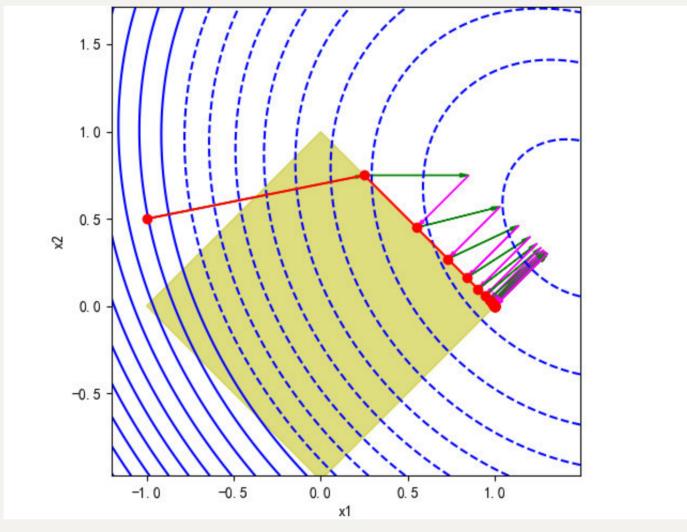
$$\mu_{2} \left[(x_{1} - 1)^{2} + x_{2}^{2} - 1 \right] = 0$$

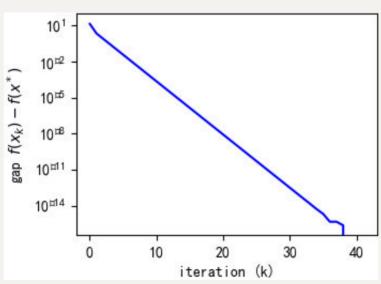
: the point $(x_1,x_2)=(1-rac{\sqrt{2}}{2},1-rac{\sqrt{2}}{2})$ is the optimal point.

Question 3

Log

t = 1 number of iterations: 41 solution: [9.99999999e-01 <u>1.00256220e-09</u>] value: 1.5





Question 4

(1.)
$$\mathcal{L}(x_1, x_2, x_3, \lambda) = e^{2x_1} + e^{x_2} + e^{x_3} + \lambda (x_1 + x_2 + x_3 - 1) \tag{14}$$

$$egin{cases} \partial_{x_1}\mathcal{L} = 2e^{2x_1} + \lambda = 0 \ \partial_{x_2}\mathcal{L} = e^{x_2} + \lambda = 0 \ \partial_{x_3}\mathcal{L} = e^{x_3} + \lambda = 0 \ x_1 + x_2 + x_3 - 1 = 0 \end{cases}$$

By solving the equations above we could get:

$$\begin{cases} x_1^* = \frac{1-2ln2}{5} \\ x_2^* = \frac{ln2+2}{5} \\ x_3^* = \frac{ln2+2}{5} \end{cases}$$
 (16)

With Lagrange multiplier and optimal value:

(2.)

KKT system for this problem is:

$$\begin{bmatrix} \nabla^2 f(\boldsymbol{x}) & \boldsymbol{A}^T \\ \boldsymbol{A} & \boldsymbol{O} \end{bmatrix} \begin{bmatrix} \boldsymbol{d} \\ \boldsymbol{\lambda} \end{bmatrix} = \begin{bmatrix} -\nabla f(\boldsymbol{x}) \\ \boldsymbol{0} \end{bmatrix}$$
(18)

Where:

$$abla f(oldsymbol{x}) = egin{pmatrix} 2e^{2x_1} \ e^{x_2} \ e^{x_3} \end{pmatrix}$$
 (19)

$$abla^2 f(m{x}) = egin{pmatrix} 4e^{2x_1} & 0 & 0 \ 0 & e^{x_2} & 0 \ 0 & 0 & e^{x_3} \end{pmatrix}$$

With A = (1, 1, 1).

By Solving equation (18), we get:

$$\begin{cases} d_1 = \frac{5}{2} \frac{e^{x_2 + x_3}}{e^{x_2 + x_3} + 4e^{2x_1 + x_3} + 4e^{2x_1 + x_2}} - \frac{1}{2} \\ d_2 = \frac{10e^{2x_1 + x_3}}{e^{x_2 + x_3} + 4e^{2x_1 + x_3} + 4e^{2x_1 + x_2}} - 1 \\ d_3 = \frac{10e^{2x_1 + x_2}}{e^{x_2 + x_3} + 4e^{2x_1 + x_3} + 4e^{2x_1 + x_2}} - 1 \end{cases}$$

$$(21)$$

(3.)

```
iteration 0: [0. 1. 0.]
iteration 1: [-0.11369186
                           0.56845929
                                       0.54523257]
iteration 2: [-0.09499115
                           0.5534111
                                       0.54158005]
iteration 3: [-0.08601059
                                       0.54002094]
                           0.54598966
iteration 4: [-0.0816069
                           0.5423022
                                       0.53930469]
iteration 5: [-0.07942602
                                       0.538962
                           0.54046402
iteration 6: [-0.07834074
                                       0.53879446]
                           0.53954628
iteration 7: [-0.07779938
                           0.53908775
                                       0.53871163]
iteration 8: [-0.07752902
                           0.53885857
                                       0.53867046]
iteration 9: [-0.07739392
                           0.53874399
                                       0.53864993]
iteration 10: [-0.07732639
                            0.53868671
                                        0.53863968]
                            0.53865807
iteration 11: [-0.07729263
                                        0.53863455]
iteration 12: [-0.07727575
                           0.53864376
                                        0.538632
iteration 13: [-0.07726731
                           0.5386366
                                        0.53863072]
iteration 14: [-0.07726309
                           0.53863302
                                        0.53863008]
iteration 15: [-0.07725887
                          0.53862944
                                        0.53862944]
iteration 16: [-0.07725887 0.53862944
                                        0.53862944]
optimal value: 4.284141440311191
```