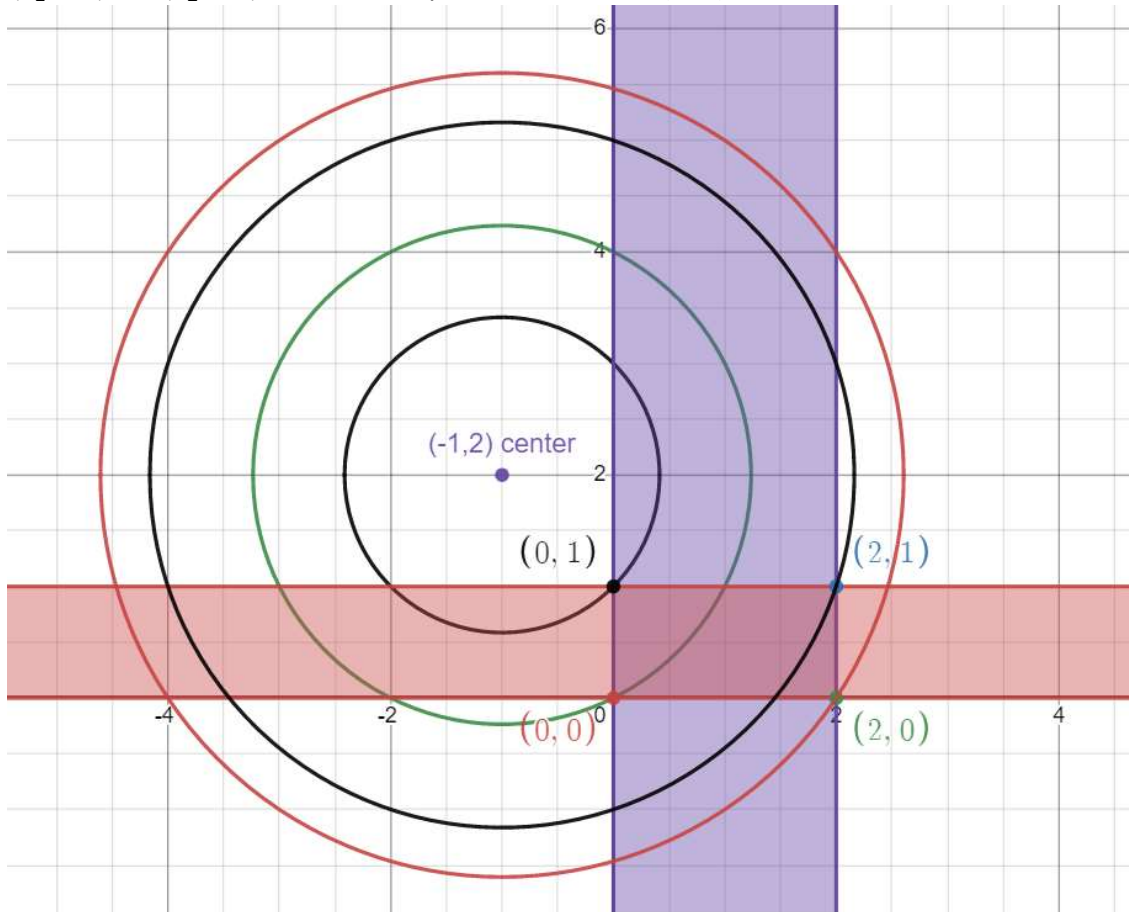


# 1. Sketch graphically the problem

$$\begin{aligned} \min_{x_1, x_2} f(x) &= (x_1 + 1)^2 + (x_2 - 2)^2 \\ \text{subjected to: } g_1 &= x_1 - 2 \leq 0, & g_3 &= -x_1 \leq 0 \\ g_2 &= x_2 - 1 \leq 0, & g_4 &= -x_2 \leq 0 \end{aligned}$$

$(x_1 + 1)^2 + (x_2 - 2)^2$  is a circle equation with center  $(-1, 2)$



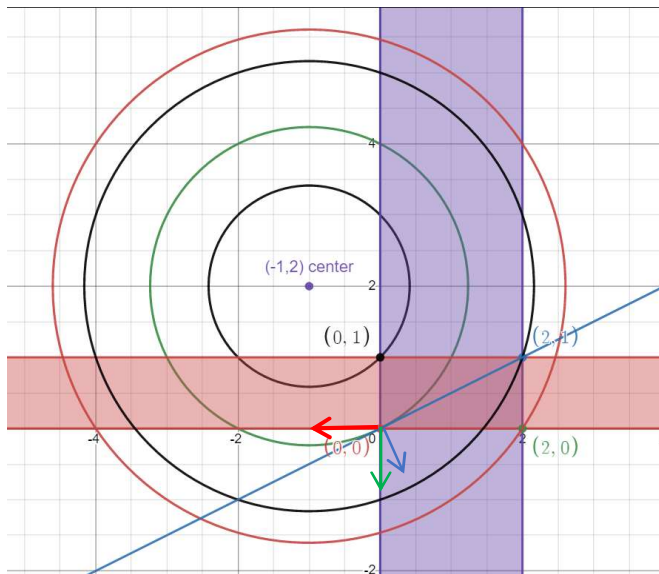
$$L = (x_1 + 1)^2 + (x_2 - 2)^2 + \mu_1(x_2 - 2) + \mu_2(x_1 - 1) + \mu_3(-x_1) + \mu_4(-x_2)$$

Conditions for  $\mu$

if $x_2 - 2 = 0$ ,	then $\mu_1 > 0$	if $x_2 - 1 < 0$ ,	then $\mu_1 = 0$
if $x_1 - 1 = 0$ ,	then $\mu_2 > 0$	if $x_1 - 2 < 0$ ,	then $\mu_2 = 0$
if $-x_1 = 0$ ,	then $\mu_3 > 0$	if $-x_1 < 0$ ,	then $\mu_3 = 0$
if $-x_2 = 0$ ,	then $\mu_4 > 0$	if $-x_2 < 0$ ,	then $\mu_4 = 0$

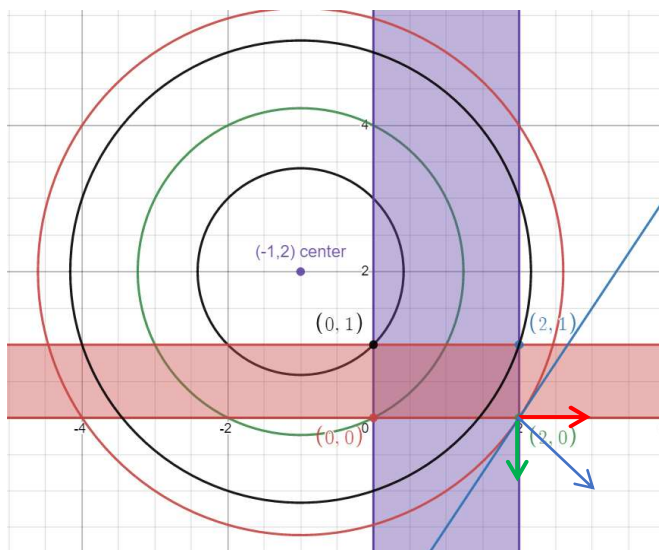
For point (0,0) the active constraints are  $g_3$  and  $g_4$

So  $\nabla g_3 = \begin{bmatrix} -1 \\ 0 \end{bmatrix}$  and  $\nabla g_4 = \begin{bmatrix} 0 \\ -1 \end{bmatrix}$



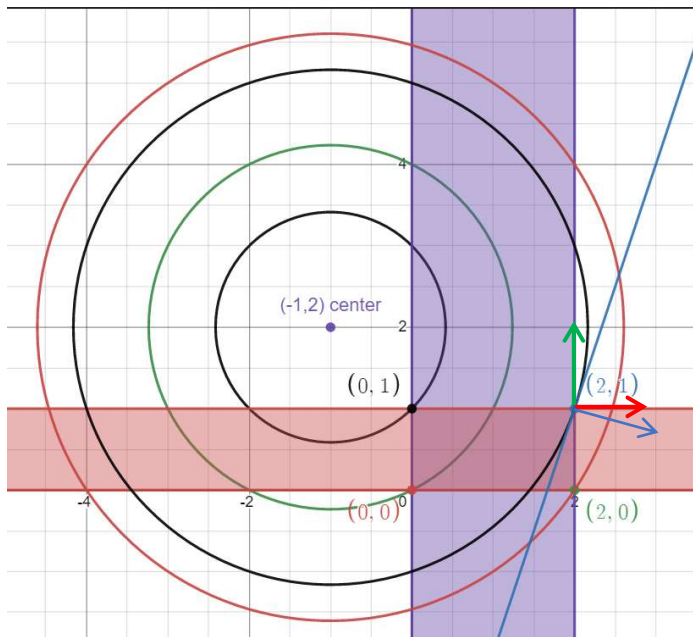
For point (2,0) the active constraints are  $g_1$  and  $g_4$

So  $\nabla g_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$  and  $\nabla g_4 = \begin{bmatrix} 0 \\ -1 \end{bmatrix}$



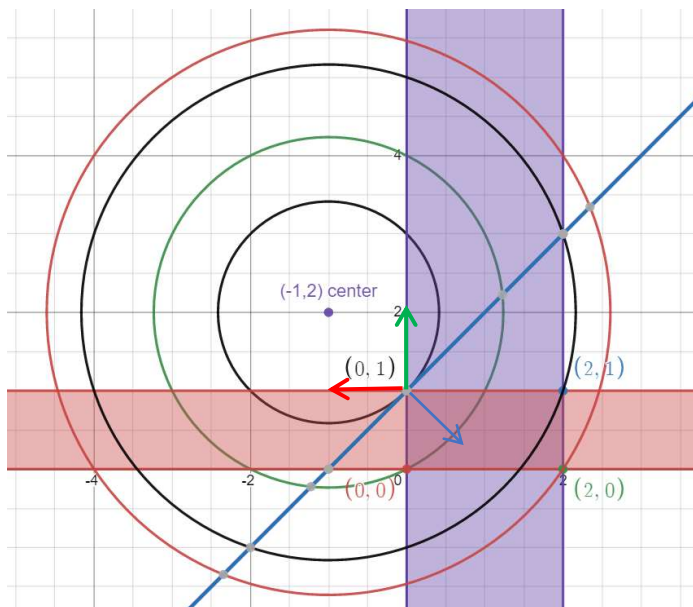
For point (2,1) the active constraints are  $g_1$  and  $g_2$

So  $\nabla g_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$  and  $\nabla g_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$



For point (0,1) the active constraints are  $g_3$  and  $g_2$

So  $\nabla g_3 = \begin{bmatrix} -1 \\ 0 \end{bmatrix}$  and  $\nabla g_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$



As we can see at point (0,1) all directions are ascent and there is no feasible descent direction,

Hence the  $x_* = (0,1)^T$  is the minimizer.

We can also check (0,1) point by Applying the KKT Necessary and Sufficient conditions.

- Necessary conditions:
  - The  $g_3$  and  $g_2$  are the active constraints mean  $\mu_3$  and  $\mu_2 > 0$  and  $\mu_1$  and  $\mu_4$  equal to 0

$$\begin{aligned} \nabla f - \mu^T \nabla g &= 0 \\ \begin{bmatrix} 2(x_1 + 1) \\ 2(x_2 - 2) \end{bmatrix} + \begin{bmatrix} -\mu_3 \\ \mu_2 \end{bmatrix} &= \begin{bmatrix} 0 \\ 0 \end{bmatrix} \\ \begin{bmatrix} 2(0 + 1) \\ 2(1 - 2) \end{bmatrix} + \begin{bmatrix} -\mu_3 \\ \mu_2 \end{bmatrix} &= \begin{bmatrix} 0 \\ 0 \end{bmatrix} \\ \begin{bmatrix} 2 - \mu_3 \\ -2 + \mu_2 \end{bmatrix} &= \begin{bmatrix} 0 \\ 0 \end{bmatrix} \end{aligned}$$

from here we have  $\mu_3 = 2$  and  $\mu_2 = 2$  that satisfied the KKT necessary conditions.

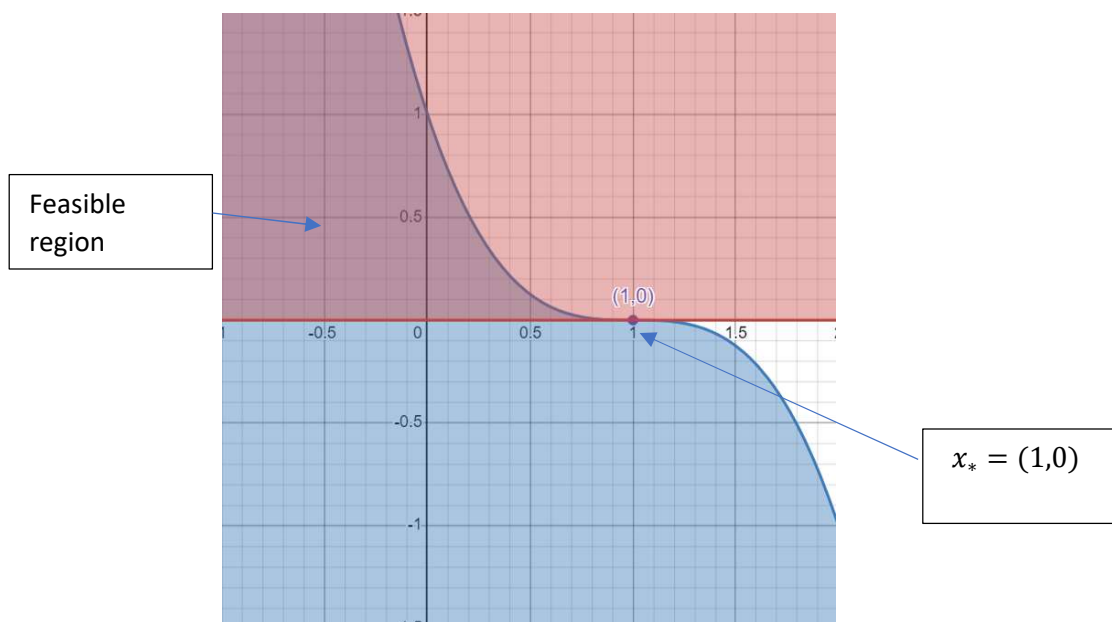
- Sufficient conditions:
  - The Hessian of Lagrangian  $= \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}$ ,  $\lambda_1 = 2$  and  $\lambda_2 = 2$

Here Hessian of Lagrangian is positive definite everywhere. Therefore,  $x_* = (0,1)^T$  is the global minimum.

## 2. Graph the problem

$$\begin{aligned} \min f - x_1 \\ \text{subject to: } g_1 = x_2 - (1 - x_1)^3 \leq 0 \text{ and } x_2 \geq 0 \end{aligned}$$

First let's convert  $x_2 \geq 0$  to  $-x_2 \leq 0$



As we can see at point  $x_* = (1,0)^T$  is the solution

Checking the KKT conditions at  $x_* = (1,0)^T$

$$L = -x_1 + \mu_1(x_2 - (1 - x_1)^3) + \mu_2(-x_2)$$

- Necessary conditions:
  - The  $g_1$  and  $g_2$  are the active constraints mean  $\mu_1$  and  $\mu_2 > 0$

$$\begin{aligned} \nabla f - \mu^T \nabla g &= 0 \\ \begin{bmatrix} -1 \\ 0 \end{bmatrix} + \mu_1 \begin{bmatrix} 3(1-x_1)^2 \\ 1 \end{bmatrix} + \mu_2 \begin{bmatrix} 0 \\ -1 \end{bmatrix} &= \begin{bmatrix} 0 \\ 0 \end{bmatrix} \end{aligned}$$

at  $x_* = (1,0)$ ,

$$\begin{bmatrix} -1 \\ 0 \end{bmatrix} + \mu_1 \begin{bmatrix} 3(1-1)^2 \\ 1 \end{bmatrix} + \mu_2 \begin{bmatrix} 0 \\ -1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} -1 \\ 0 \end{bmatrix} + \mu_1 \begin{bmatrix} 0 \\ 1 \end{bmatrix} + \mu_2 \begin{bmatrix} 0 \\ -1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\begin{aligned} -1 &= 0: \text{contradict} \\ \mu_1 - \mu_2 &= 0 \end{aligned}$$

So, the point  $x_* = (1,0)$  is not a KKT point because this is not a regular point.

### 3. Find a local solution to the problem

$$\max f = x_1 x_2 + x_2 x_3 + x_1 x_3$$

$$\text{subjected to } h = x_1 + x_2 + x_3 - 3 = 0$$

You can use either the reduced gradient or the Lagrangian method.

- Solving it using Lagrangian Method:

$$\begin{aligned} L &= -f + \lambda h \\ L &= -(x_1 x_2 + x_2 x_3 + x_1 x_3) + \lambda(x_1 + x_2 + x_3 - 3) \\ \nabla_x L &= \begin{bmatrix} -x_2 - x_3 + \lambda \\ -x_1 - x_2 + \lambda \\ -x_1 - x_3 + \lambda \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \\ \nabla_\lambda L &= x_1 + x_2 + x_3 - 3 = 0 \end{aligned}$$

We have 4 unknown and 4 equations so by solving the system of linear equation

$$x_1 = 1, \quad x_2 = 1, \quad x_3 = 1, \quad \lambda = 2$$

- Check Sufficient conditions:

- The Hessian of Lagrangian  $L_{xx} = \begin{bmatrix} 0 & -1 & -1 \\ -1 & 0 & -1 \\ -1 & -1 & 0 \end{bmatrix}$
- The eigen value of Hessian of Lagrangian is  $\lambda_1 = -2, \lambda_2 = 1, \lambda_3 = 1$ .
  - As we can see all the eigen value of Hessian are not positive

But if we check the second order condition which is,  $dx^T L_{xx} dx$

$\partial x L_{xx} \partial x$  : Second order perturbations

$$\begin{aligned} \partial x^T L_{xx} \partial x &= [\partial x_1 \ \partial x_2 \ \partial x_3] \begin{bmatrix} 0 & -1 & -1 \\ -1 & 0 & -1 \\ -1 & -1 & 0 \end{bmatrix} \begin{bmatrix} \partial x_1 \\ \partial x_2 \\ \partial x_3 \end{bmatrix} \\ &= -2\partial x_1 \partial x_2 - 2\partial x_1 \partial x_3 - 2\partial x_2 \partial x_3 \end{aligned}$$

We want the  $\partial x$  to be feasible so the feasible perturbation is such that  $\frac{\partial h}{\partial x} \partial x = 0$

$$\left[ \frac{\partial h}{\partial x_1}, \frac{\partial h}{\partial x_2}, \frac{\partial h}{\partial x_3} \right] \begin{bmatrix} \partial x_1 \\ \partial x_2 \\ \partial x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\begin{aligned} [1, 1, 1] \begin{bmatrix} \partial x_1 \\ \partial x_2 \\ \partial x_3 \end{bmatrix} &= 0 \rightarrow \partial x_1 + \partial x_2 + \partial x_3 = 0 \\ \partial x_1 &= -\partial x_2 - \partial x_3 \end{aligned}$$

So,

$$\begin{aligned} &= -2(-\partial x_2 - \partial x_3)\partial x_2 - 2(-\partial x_2 - \partial x_3)\partial x_3 - 2\partial x_2 \partial x_3 \\ &= 2(\partial x_2^2 + \partial x_2 \partial x_3 + \partial x_3^2) \rightarrow 2\left(\partial x_2 + \frac{1}{2}\partial x_3\right)^2 + \frac{3}{4}\partial x_3^2 \geq 0 \end{aligned}$$

Further, for  $\partial x^T L_{xx} \partial x$  to be 0,  $\partial x_2$  and  $\partial x_3$  must be 0, if so the  $\partial x_1$  is also 0. Which mean  $\partial x = 0$  which is not a perturbation. Therefore,  $\partial x^T L_{xx} \partial x > 0$  for any non-zero feasible perturbations

So  $x_{1*} = x_{2*} = x_{3*} = 1$  is global maximum to the original problem and if we plug this value in the main  $f$  so,  $f_* = 3$

#### 4. Find the solution to

$$\min f = x_1^2 + x_2^2 + x_3^2$$

$$\begin{aligned} \text{subjected to } h_1 &= \frac{x_1^2}{4} + \frac{x_2^2}{5} + \frac{x_3^2}{25} - 1 = 0 \\ \text{and } h_2 &= x_1 + x_2 - x_3 = 0 \end{aligned}$$

by implementing the generalized reduced gradient algorithm.

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