

1) a)

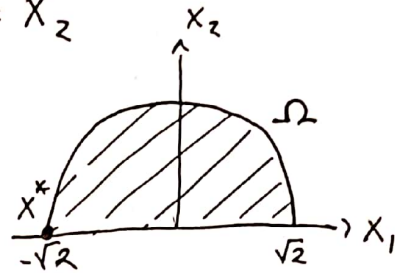
$$\min x_1 + 2x_2 \quad \text{s.t.} \quad 2 - x_1^2 - x_2^2 \geq 0, \quad x_2 \geq 0$$

$$c_1(x) = 2 - x_1^2 - x_2^2, \quad c_2(x) = x_2$$

$$\mathcal{L}(x, \lambda) = f(x) - \lambda_1 c_1 - \lambda_2 c_2$$

It is trivial to see that the optimal point is

$$x^* = (-\sqrt{2}, 0)^T$$



$$\textcircled{b} \quad \mathcal{L}(x, \lambda) = x_1 + 2x_2 - \lambda_1 (2 - x_1^2 - x_2^2) - \lambda_2 x_2$$

$$\nabla_x \mathcal{L}(x, \lambda) = \begin{bmatrix} 1 + 2\lambda_1 x_1 \\ 2 + 2\lambda_1 x_2 - \lambda_2 \end{bmatrix} \stackrel{!}{=} 0$$

$$\nabla_x \mathcal{L}(x^*, \lambda^*) = \begin{bmatrix} 1 - 2\sqrt{2} \lambda_1^* \\ 2 - \lambda_2^* \end{bmatrix} = 0$$

$$\Rightarrow \underline{\lambda^* = \begin{bmatrix} 1/2\sqrt{2} \\ 2 \end{bmatrix} > 0}$$

All the constraints are active in  $x^*$ ,

$$\text{i.e. } c_1^* = 2 - x_1^{*2} - x_2^{*2} = 2 - \sqrt{2}^2 = 0$$

$$c_2(x^*) = x_2^* = 0$$

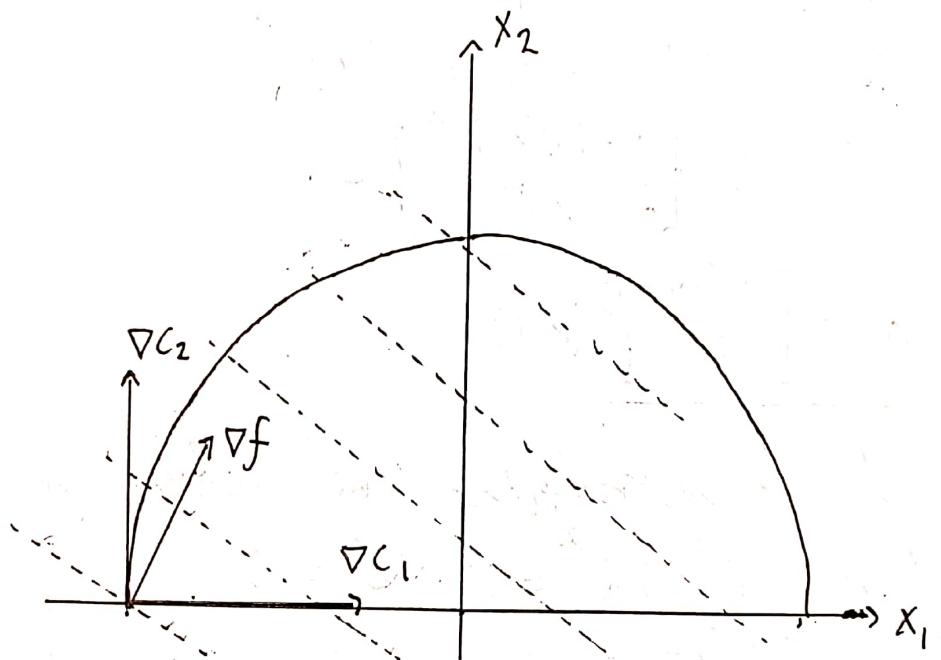
Since  $c_i(x^*) = 0$  it also follows that

$$\lambda_i^* c_i(x^*) = 0 \quad \text{for } i \in [1, 2],$$

so all the KKT conditions are satisfied.

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$$\nabla f(x^*) = \begin{bmatrix} 1 \\ 2 \end{bmatrix}, \quad \nabla C_1(x^*) = \begin{bmatrix} 2\sqrt{2} \\ 0 \end{bmatrix}, \quad \nabla C_2(x^*) = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$



② The Lagrange multipliers has to be positive, as we want the gradient of the objective function and the gradient of the constraints to point in the same direction, not opposite.

- © The feasible set is clearly convex, and since  $f(x)$  is linear it is also convex.  
 viz the problem is convex.

[2]  $\min 2x_1 + x_2 \quad \text{s.t.} \quad x_1^2 + x_2^2 - 2 = 0$

①  $\mathcal{L}(x, \lambda) = 2x_1 + x_2 - \lambda(x_1^2 + x_2^2 - 2)$

$\nabla_x \mathcal{L}(x^*, \lambda^*) = \begin{bmatrix} 2 - 2\lambda^*x_1^* \\ 1 - 2\lambda^*x_2^* \end{bmatrix} = 0$

So an extreme point has the form  $(1/\lambda^*, 1/2\lambda^*)^T$ .

We also have that

$$x_1^{*2} + x_2^{*2} - 2 = 1/\lambda^{*2} + 1/4\lambda^{*2} - 2 = 0$$

$$\Rightarrow \lambda^* = \pm \sqrt{\frac{5}{8}}$$

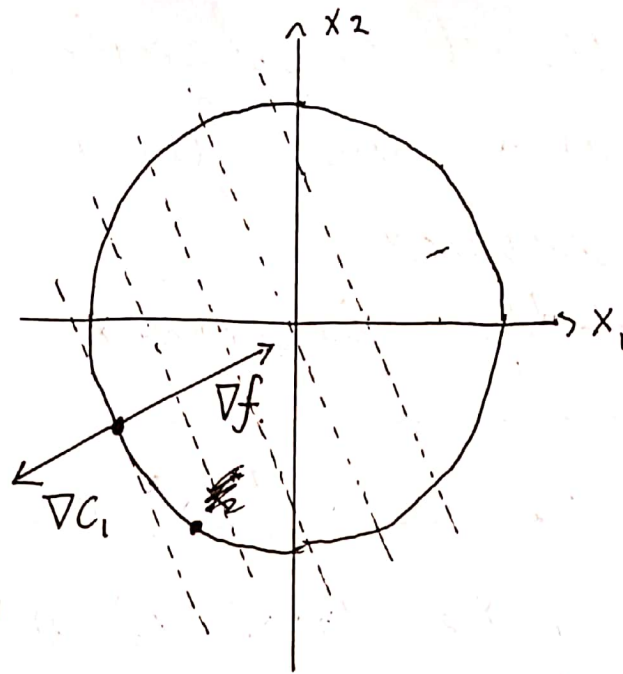
- Which means the extreme points are  $(\sqrt{\frac{8}{5}}, \sqrt{\frac{2}{5}})$  and  $(-\sqrt{\frac{8}{5}}, -\sqrt{\frac{2}{5}})$ .

The solution is obviously  $(-\sqrt{\frac{8}{5}}, -\sqrt{\frac{2}{5}})$ .

- ⑥ We have already shown that the stationary condition  $\nabla_x \mathcal{L}(x^*, \lambda^*) = 0$  and the feasibility condition  $c_1(x^*) = 0$  are met in the two extreme points.

- The complementary condition is also met, as  $c_1(x^*) = 0 \Rightarrow \lambda_1^* c_1(x^*) = 0$ . The two extreme points are the max and min in  $\Omega$ , so KKT should be true in both cases.

(c)



$$\nabla f(x^*) = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$$

$$\nabla C_1(x^*) = \begin{bmatrix} -2\sqrt{\frac{8}{5}} \\ -2\sqrt{\frac{2}{5}} \end{bmatrix} = \begin{bmatrix} -4\sqrt{2/5} \\ -2\sqrt{2/5} \end{bmatrix}$$

(d) We found previously that

$$\underline{\underline{\lambda^* = -\sqrt{\frac{5}{8}}}}$$

Since we have an equality constraint the sign is not important and therefore not a part of the KKT conditions, so it is consistent with KKT. (Only multipliers associated with an inequality constraint need to be nonnegative).



© The second order condition states that:

$$w^T \nabla_{xx}^2 \mathcal{L}(x^*, \lambda^*) w \geq 0 \quad \forall w \in C(x^*, \lambda^*)$$

Let  $w = \begin{bmatrix} w_1 \\ w_2 \end{bmatrix}$ .

$$\begin{aligned} w^T \nabla_{xx}^2 \mathcal{L}(x^*, \lambda^*) w &= [w_1, w_2] \cdot \begin{bmatrix} -2\lambda^* & 0 \\ 0 & -2\lambda^* \end{bmatrix} \cdot \begin{bmatrix} w_1 \\ w_2 \end{bmatrix} \\ &= -2\lambda^* (w_1^2 + w_2^2) \end{aligned}$$

We observe that the sign of this product is only dependent on  $\lambda^*$ .

So for the extreme point  $(2\sqrt{\frac{2}{5}}, \sqrt{\frac{2}{5}})$  we have that  $\lambda^* > 0$  so the 2. order condition is not satisfied. This is to be expected, as this point is the maxima.

The other extreme point  $(-2\sqrt{\frac{2}{5}}, -\sqrt{\frac{2}{5}})$  has  $\lambda^* < 0$  so the 2. order condition is satisfied, and this point is therefore the optimal point.

Ⓕ While the objective function is convex, the feasible set is not. Any two points on the circle  $C_1(x) = 0$  which are not identical will leave the circle, so it is not a convex problem.

$$\boxed{3} \quad \min f(x) = -2x_1 + x_2 \quad \text{s.t.} \quad \begin{aligned} C_1(x) &= (1-x_1)^3 - x_2 \geq 0, \\ C_2(x) &= x_2 + 0,25x_1^2 - 1 \geq 0 \end{aligned}$$

$$\textcircled{a} \quad x^* = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

$$\nabla C_1(x^*) = \begin{bmatrix} -3(1-x_1^*)^2 \\ -1 \end{bmatrix} = \begin{bmatrix} -3 \\ -1 \end{bmatrix}$$

$$\nabla C_2(x^*) = \begin{bmatrix} 0,5x_1^* \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

$\nabla C_1(x^*)$  and  $\nabla C_2(x^*)$  are clearly linearly independent, so LICQ holds in  $x^*$ .

$\textcircled{b}$

$$| \quad \mathcal{L}(x, \lambda) = -2x_1 + x_2 - \lambda_1((1-x_1)^3 - x_2) - \lambda_2(x_2 + 0,25x_1^2 - 1)$$

$$\nabla_x \mathcal{L}(x, \lambda) = \begin{bmatrix} -2 + 3\lambda_1(1-x_1)^2 - 0,5\lambda_2x_1 \\ 1 + \lambda_1 - \lambda_2 \end{bmatrix}$$

$$\nabla_x \mathcal{L}(x^*, \lambda^*) = \begin{bmatrix} -2 + 3\lambda_1^* \\ 1 + \lambda_1^* - \lambda_2^* \end{bmatrix} = 0$$

$$\Rightarrow \underline{\lambda^* = \begin{bmatrix} 2/3 \\ 5/3 \end{bmatrix} \geq 0} \quad \begin{aligned} C_1(x^*) &= 1-1=0, \\ C_2(x^*) &= 1-1=0 \end{aligned}$$

Since all ~~constraints~~ constraints are active the complementary conditions are also satisfied. Thus KKT is satisfied.

$$\textcircled{d} \quad \nabla_{xx}^2 \mathcal{L}(x^*, \lambda^*) = \begin{bmatrix} -6\lambda_1^*(1-x_1^*) - 0.5\lambda_2^* & 0 \\ 0 & 0 \end{bmatrix}$$

$$= \begin{bmatrix} -\frac{29}{6} & 0 \\ 0 & 0 \end{bmatrix}$$

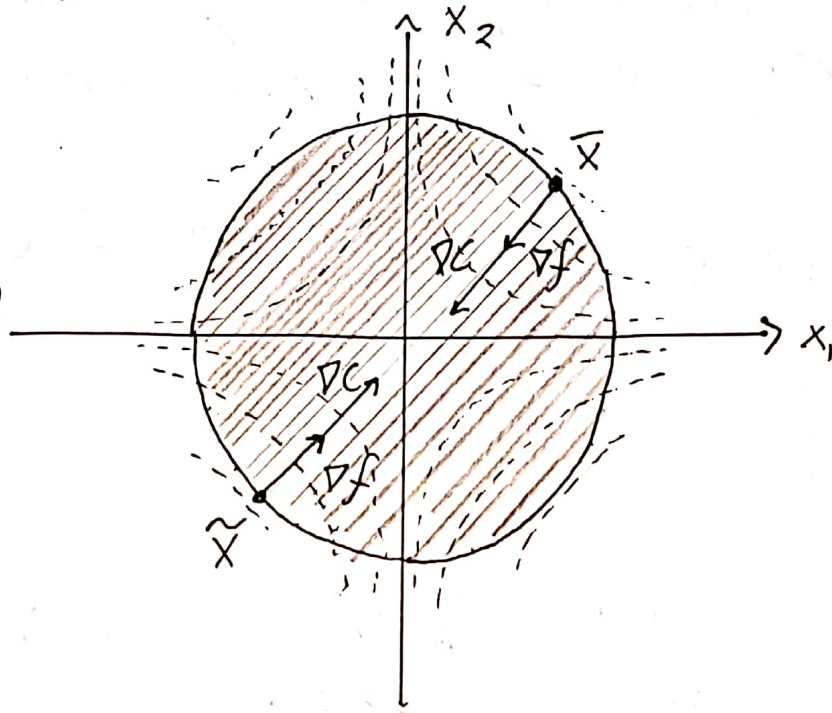
$$\textcircled{e} \quad w^T \nabla_{xx}^2 \mathcal{L}(x^*, \lambda^*) = -\frac{29}{6} w_1^2 \geq 0$$

The above equation has to hold for  $\nabla C_1(x^*)^T w = 0$  and  $\nabla C_2(x^*)^T w = 0$ , which is only true for  $w = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$ , so

$w^T \nabla_{xx}^2 \mathcal{L}(x^*, \lambda^*) = 0$ , and therefore the 2. order necessary conditions are satisfied, but the sufficient conditions are not.

4]  $\min f(x) = -x_1 x_2 \quad \text{s.t.} \quad C_1(x) = 1 - x_1^2 - x_2^2 \geq 0$

$$\begin{aligned} \nabla \bar{f} &= \begin{bmatrix} -1/\sqrt{2} \\ -1/\sqrt{2} \end{bmatrix} \\ \nabla \bar{C} &= \begin{bmatrix} \sqrt{2} \\ -\sqrt{2} \end{bmatrix} \\ \nabla \tilde{f} &= \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix} \\ \nabla \tilde{C} &= \begin{bmatrix} \sqrt{2} \\ \sqrt{2} \end{bmatrix} \end{aligned}$$



$$\mathcal{L}(x, \lambda) = -x_1 x_2 - \lambda(1 - x_1^2 - x_2^2)$$

$$\nabla_x \mathcal{L}(x, \lambda) = \begin{bmatrix} -x_2 + 2\lambda x_1 \\ -x_1 + 2\lambda x_2 \end{bmatrix} = 0$$

$$\Rightarrow x_1 = 2\lambda x_2, \quad x_2 = 2\lambda x_1 = (2\lambda)^2 x_2$$

$$\Rightarrow 4\lambda^2 = 1$$

$$2\lambda = 1 \quad (\text{as } \lambda^* \geq 0)$$

Assume that  $C_1$  is active.

It is trivial to see graphically that this has to be the case.

$$\rightarrow 1 = x_1^2 + x_2^2 = (1 + 4\lambda^2) x_2^2 = 2x_2^2$$

$$\Rightarrow x_2 = \pm \frac{1}{\sqrt{2}},$$

$$x_1 = 2\lambda x_2 = x_2 = \pm \frac{1}{\sqrt{2}}$$

So the two solutions are  $x^* = \pm(\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}})$ .