

①  $f(x_1, x_2) = 2x_1^2 - 4x_1x_2 + 1.5x_2^2 + x_2$

$$\nabla f = \begin{bmatrix} \frac{\partial f}{\partial x_1} \\ \frac{\partial f}{\partial x_2} \end{bmatrix} = \begin{bmatrix} 4x_1 - 4x_2 \\ -4x_1 + 3x_2 + 1 \end{bmatrix}$$

$$H = \begin{bmatrix} \frac{\partial^2 f}{\partial x_1^2} & \frac{\partial^2 f}{\partial x_1 \partial x_2} \\ \frac{\partial^2 f}{\partial x_2 \partial x_1} & \frac{\partial^2 f}{\partial x_2^2} \end{bmatrix} = \begin{bmatrix} 4 & -4 \\ -4 & 3 \end{bmatrix}$$

Solving the gradient for the stationary point.

$$\left. \begin{array}{l} 4x_1 - 4x_2 = 0 \\ -4x_1 + 3x_2 + 1 = 0 \end{array} \right\} \begin{array}{l} \text{Solving the two equations} \\ \text{we get } x_1 = 1 \text{ \& } x_2 = 1 \end{array}$$

Taylor's series at the saddle point:

$$f(x_1, x_2) \approx f(x_0) + \nabla f(x_0)^T \cdot (x - x_0) + \frac{1}{2} (x - x_0)^T H(x_0) \cdot (x - x_0)$$

$$x_0 = \begin{bmatrix} 1 \\ 1 \end{bmatrix} \quad \therefore f(x_0) = 0.5$$



$$\nabla f(x_0) = 0$$

$$x - x_0 = \begin{bmatrix} x_1 - 1 \\ x_2 - 1 \end{bmatrix} = \begin{bmatrix} dx_1 \\ dx_2 \end{bmatrix}$$

$$f(x) = 0.5 + \frac{1}{2} \begin{bmatrix} dx_1 & dx_2 \end{bmatrix} \begin{bmatrix} 4 & -4 \\ -4 & 3 \end{bmatrix} \begin{bmatrix} dx_1 \\ dx_2 \end{bmatrix}$$

$$f(x_1, x_2) = 0.5 + \frac{1}{2} \left[ 4 dx_1^2 - 8 dx_1 dx_2 + 3 dx_2^2 \right]$$

$$f(x_1, x_2) - 0.5 = \frac{1}{2} \left[ 4 dx_1^2 - 8 dx_1 dx_2 + 3 dx_2^2 \right]$$

This can be factorized as

$$f(x_1, x_2) - 0.5 = \frac{1}{2} (2 dx_1 - dx_2)(2 dx_1 - 3 dx_2)$$

To get the down slopes we can show

$$2 dx_1 - dx_2 < 0 \text{ \& } 2 dx_1 - 3 dx_2 > 0$$

or

$$2 dx_1 - dx_2 > 0 \text{ \& } 2 dx_1 - 3 dx_2 < 0.$$



Q2) we can make the objective function as.

$$d^2 = (x+1)^2 + (y)^2 + (z-1)^2$$

we use the constraint  $\rightarrow x + 2y + 3z = 1$

$$\therefore z = \frac{1-x-2y}{3}$$

$$d^2 = (x+1)^2 + y^2 + \left(\frac{1-x-2y-3}{3}\right)^2$$

$$\therefore d^2 = x^2 + 2x + 1 + y^2 + \frac{x^2 + 4xy + 4x + 8y + 4y^2 + 4}{9}$$

$$\therefore f(x,y) = d^2 = (10x^2 + 22x + 13y^2 + 8y + 4xy + 13)/9$$

$$\nabla f(x,y) = \begin{bmatrix} \partial f / \partial x \\ \partial f / \partial y \end{bmatrix} = \begin{bmatrix} 20x + 22 + 4y/9 \\ 26y + 8 + 4x/9 \end{bmatrix}$$

$$H = \begin{bmatrix} 20/9 & 4/9 \\ 4/9 & 26/9 \end{bmatrix}$$

$\rightarrow$  Hence its convex!

To find the  $x$  &  $y$  values when gradient is zero.

$$20x + 4y = -22$$

$$4x + 26y = -8$$

solving for unknown  $x$  &  $y$  we get

$$\begin{cases} x = -15/14 \\ y = -1/7 \end{cases}$$

when we put  $x$  &  $y$  back into equation for  $z$  we get

$$z = 11/14$$



③

$$H = \left\{ \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \in \mathbb{R}^n \mid a_1 x_1 + \dots + a_n x_n = c \right\}$$

where  $a_1, \dots, a_n \neq 0$  and  $c \in \mathbb{R}$

we need to show

$$\forall x, y \in H \quad \lambda x + (1-\lambda)y \in H \quad 0 \leq \lambda \leq 1$$

$$\lambda x + (1-\lambda)y = \begin{bmatrix} \lambda x_1 + (1-\lambda)y_1 \\ \vdots \\ \lambda x_n + (1-\lambda)y_n \end{bmatrix}$$

we can write it as

$$\begin{aligned} a_1 x_1 + \dots + a_n x_n &= c \\ a_1 y_1 + \dots + a_n y_n &= c \end{aligned}$$

$\therefore$  for all  $\lambda \in [0, 1]$

$$\lambda a_1 x_1 + \dots + \lambda a_n x_n = \lambda c$$

$$(1-\lambda)a_1 y_1 + \dots + (1-\lambda)a_n y_n = (1-\lambda)c$$

summing up the above equation.

$$a_1 (\lambda x_1 + (1-\lambda)y_1) + \dots + a_n (\lambda x_n + (1-\lambda)y_n) = \lambda c + (1-\lambda)c$$

$$a_1 (\lambda x_1 + (1-\lambda)y_1) + \dots + a_n (\lambda x_n + (1-\lambda)y_n) = c$$

$$\Rightarrow \boxed{\lambda x + (1-\lambda)y \in H}$$



4 a) Show that the problem is convex.

$$h = I_t / I$$

$$h(p) = I_t / a^T p$$

$$\nabla h = \frac{I_t}{-(a^T p)^2} a$$

$$H = \frac{2 I_t a \cdot a^T}{(a^T p)^3} \} \rightarrow \text{since the Hessian } > 0 \text{ we can say that the problem is } \text{not} \text{ convex}$$

b) The problem will be convex as the function will become a combination of different half spaces and hence its convex.

$$\text{ex: } p_1 + p_2 + \dots + p_n \leq p^*$$

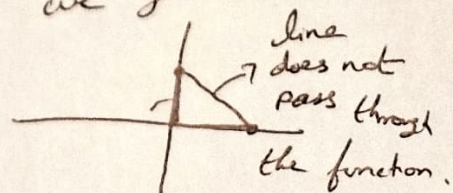
$$p_2 + \dots + p_n < p^*$$

Although the function becomes convex, it is not necessary for it to have a unique solution

c) The problem will not be convex as the function created will not follow the rule of  $\boxed{\lambda x_1 + (1-\lambda) x_2 \in S}$

for a convex set  $S$ .

For example, we have two lamps + two cases when one is on & the other is off & vice versa then we get ~~two~~ function that is non convex





$$5) \quad C^*(y) = \max_x \{ xy - c(x) \}$$

Show that  $C^*(y)$  is convex with respect to  $y$ .

We can check the second differential of the function

$$\frac{dC^*(y)}{dy} = x$$

$$\frac{d^2 C^*(y)}{dy^2} = 0 \rightarrow \text{Since the second differential is zero we can say that the function is convex.}$$

max of a convex function is also convex.