

## 36 Lecture 36

[36.1] Example Solve  $\min f(x) = (x_1 - 1)^2 + x_2 - 2$   
s.t.  $g(x) \equiv 2 - x_1 - x_2 \geq 0, h(x) \equiv x_2 - x_1 - 1 = 0. \rightarrow T.$

• constraints are linear. so KTCQ1 holds at each feasible pt.

• By KTNC, if  $a$  is a pt of <sup>local</sup> minimum, then it must be a KT point.

• Find KT points:  $\exists \lambda > 0, w \in \mathbb{R}$

$$\begin{bmatrix} 2x_1 - 2 \\ 1 \end{bmatrix} = \lambda \begin{bmatrix} -1 \\ -1 \end{bmatrix} + w \begin{bmatrix} -1 \\ 1 \end{bmatrix}, \lambda(2 - x_1 - x_2) > 0, a \in T$$

• check if  $\lambda = 0$  is possible. if  $\lambda = 0$ , then  $w = 1$ . so  $x_1 = .5, x_2 = 1.5$

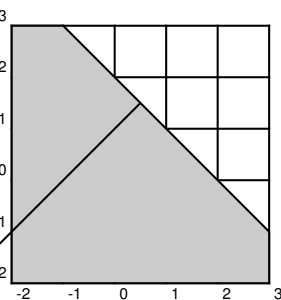
• check if  $\lambda > 0$ . if  $\lambda > 0$ , then  $x_1 + x_2 = 2, x_2 - x_1 = 1 \Rightarrow x_2 = 1.5, x_1 = .5$

As  $\lambda > 0$ , we put  $w > 1$ .  $2x_1 - 2 = -\lambda - w < -1 \Rightarrow 2x_1 < 1 \Rightarrow \Leftarrow$

• so  $(.5, 1.5)$  is the only KT point with  $\lambda = 0, w = 1$ .

This ray is our feasible region.  
We do not need it though.

$$\begin{aligned} \mathcal{D}_+(a) &= \{d \mid \begin{bmatrix} -1 & -1 \end{bmatrix} \begin{bmatrix} d_1 \\ d_2 \end{bmatrix} > 0 \\ &\quad \begin{bmatrix} -1 & 1 \end{bmatrix} \begin{bmatrix} d_1 \\ d_2 \end{bmatrix} = 0 \} \\ &= \{d \mid d_1 + d_2 < 0, d_1 = d_2\} \\ &= \{ \begin{bmatrix} -t \\ -t \end{bmatrix} \mid t > 0 \} \end{aligned}$$



• Apply KTS  $H_L(a, \lambda, w)$   
 $= \begin{bmatrix} 2 & 0 \\ 0 & 0 \end{bmatrix}$  not pd  
psd in a nbhd  
so it is a local min.

◦ Constraints are linear. So KTCQ1 holds at each feasible point. So by KTNC, each point of local minimum is a KT point. Denote the feasible region by  $T$ .

◦ We find KT points:

$$\begin{bmatrix} 2(x_1 - 1) \\ 1 \end{bmatrix} = \lambda \begin{bmatrix} -1 \\ -1 \end{bmatrix} + w \begin{bmatrix} -1 \\ 1 \end{bmatrix}, \quad \lambda(2 - x_1 - x_2) = 0, \quad x \in T.$$

so by KTS  $a$ ,  $a$  is a st local min.

◦ Check if  $\lambda = 0$ .

If  $\lambda = 0$ , then  $w = 1$ . Equating the first row we get  $x_1 = .5$ . Applying  $h(x)$ , we get  $x_2 = 1.5$ . Thus  $(.5, 1.5)$  being feasible, is a KT point with  $\lambda = 0, w = 1$ .

◦ Check if  $\lambda > 0$  gives more KT points.

If  $\lambda > 0$ , then  $x_1 + x_2 = 2$ . Using  $h(x)$ , we get  $x_2 = 1.5$  and  $x_1 = .5$ . But as  $w > 1$ , we must have  $2x_1 < 1$ , not possible. So we do not get any more KT points.

◦ So  $a = (.5, 1.5)$  is the only KT point with  $\lambda = 0$ ,  $w = 1$ . Use KTSC.

We have  $H_L = \begin{bmatrix} 2 & 0 \\ 0 & 0 \end{bmatrix}$ . It is psd in a neighborhood of  $a$ . So by KTSC c),  $a$  is a local minimum.

More information: We have  $g$  is active with  $\lambda = 0$ . So  $A_+(a) = \emptyset$ . So

$$\mathcal{D}_+(a) = \mathcal{D}(a) = \{d \mid d_1 + d_2 \leq 0, d_1 = d_2\}.$$

Take an arbitrary nonzero  $d = \begin{bmatrix} -t \\ -t \end{bmatrix} \in \mathcal{D}_+(a)$ . Then  $t > 0$ . So we have  $d^T H(L)_a d = 2t^2 > 0$ . By KTSC a),  $a$  is a strict local minimum.

[36.2] **Example** (One dimensional example.) Consider  $\min \frac{x^2}{x \geq 5, x \leq 6}$  and  $\max \frac{x^2}{x \geq 5, x \leq 6}$ .

Find the KT points for both the problems apply KTSC.

*Answer.* min. constraints lin  $\Rightarrow$  KTSC holds  $\Rightarrow$  pts of local min must be KT pts. Find KT points. Write  $g_1 = x - 5 \geq 0$ ,  $g_2 = 6 - x \geq 0$

KT cond  $\exists \lambda_i \geq 0$  s.t.  
 $2x = \lambda_1 - \lambda_2$ ,  $\lambda_1(x - 5) = 0$ ,  $\lambda_2(6 - x) = 0$ ,  $x \in T$

• If  $\lambda_1 = 0$  possible. If  $\lambda_1 = 0$ , then  $2x = -\lambda_2$ ,  $\Rightarrow$  not possible.  
 • So  $\lambda_1 > 0$ . So  $x = 5$ . So  $\lambda_2 = 0$ . So  $\lambda_1 = 10$ . So  $x = 5$  is a KT pt with  $\lambda_1 = 10, \lambda_2 = 0$ .

Apply KTSC  $H_L(5) = [2]$  pd.  $\Rightarrow$  By KTSC b), 5 is a st local min.

For max  $x^2$ , it is min  $-x^2$ . You get  $x = 6$  is a KT pt with  $\lambda_2 = 12, \lambda_1 = 0$ .  $H_L(6) = [-2]$  not pd, not psd.  $A_+(6) = \{2\}$   
 $\mathcal{D}_+(6) = \{d \mid -d = 0\} = \{0\}$ . So for every nonzero  $d \in \mathcal{D}_+(6)$  we have  $d^T H_L d > 0$ .

a) First consider the minimization problem. We have  $L = x^2 - \lambda_1(x - 5) - \lambda_2(6 - x)$ . It is a st local min.

b) Constraints are linear. So ktcq1 holds at all feasible points. Any local minimum must be a kt point.

c) Find kt points:

$$2x = \lambda_1 - \lambda_2.$$

If  $\lambda_2 > 0$ , then  $x = 6$ . As  $\lambda_1 g_1(a) = 0$ , we must have  $\lambda_1 = 0$ . But then  $12 = 0 - \lambda_2$  is not possible.

So  $\lambda_2 = 0$ . If  $\lambda_1 > 0$ , then  $x = 5$ . Notice that  $x = 5$  is a kt point with  $\lambda_1 = 10$  and  $\lambda_2 = 0$ . Notice that  $H_L = [2]$  is pd. Hence the point  $x = 5$  is a strict local minimum by ktsc.

A) Consider the maximization problem, that is, minimizing  $-x^2$ . Approaching similarly, we get  $a = 6$  is the only KT point with  $\lambda_1 = 0$  and  $\lambda_2 = 12$ .

B) Use KTSC: Here  $H_L = -2$ . So we cannot use parts b) and c) of KTSC.

C) However, we can use part a). Note that  $A_+(6) = \{2\}$  and  $\mathcal{D}_+(6) = \{0\}$ . Hence, for each nonzero  $d \in \mathcal{D}_+(6)$  we have  $d^t H(L) d > 0$ . So  $x = 6$  satisfies KTSC part a), to be a point of strict local minimum for  $-x^2$ .

D) So  $a = 6$  is a strict local maximum for  $f(x) = x^2$ .

E) You can conclude it to be absolute maximum, as the function being continuous on a compact set must attain its absolute maximum. That point has to be a KT point. But there is only one KT point. So  $a = 6$  is also an absolute maximum.

## Exercises

[36.3] **NoPen** Let  $a$  be any feasible point for  $\max \frac{f(x)}{\text{s.t. } \begin{array}{l} g_i(x) \geq 0, i = 1, \dots, m, \\ h_j(x) = 0, j = 1, \dots, p. \end{array}}$

Suppose that  $\mathcal{D}(a) \cap \{d \mid \nabla f(a)^t d > 0\} = \emptyset$ . Is it necessary that  $a$  is a KT point?

[36.4] **Exercise(M)** Let  $a$  be a local minimum for

$$\min \frac{f(x)}{\text{s.t. } \begin{array}{l} g_i(x) \geq 0, i = 1, \dots, m, \quad h_j(x) = 0, j = 1, \dots, p, \quad x_k \geq 0, k = 1, \dots, n. \end{array}}$$

a) Assume that  $Z(a) = \emptyset$ . Let

$$L(x, \lambda, w) = f(x) - \sum_{i=1}^m \lambda_i g_i(x) - \sum_{j=1}^p w_j h_j(x).$$

Prove that  $\exists \lambda_i \geq 0, i = 1, \dots, m, w_j, j = 1, \dots, p$  such that the following KT conditions are satisfied

$$\nabla L(a, \lambda, w) \geq 0, \quad \lambda_i g_i(a) = 0, \forall i = 1, \dots, m, \quad a^t \nabla L(a, \lambda, w) = 0.$$

b) Is the converse of a) true?

c) Is  $a = (0, \sqrt{2}, \sqrt{2})$  a KT point of

$$\min \frac{f = x_1^3 - 6x_1^2 + 11x_1 + x_3}{\text{s.t. } \begin{array}{l} g_1 \equiv -x_1^2 - x_2^2 + x_3^2 \geq 0, \quad g_2 \equiv x_1^2 + x_2^2 + x_3^2 - 4 \geq 0, \quad g_3 \equiv 5 - x_3 \geq 0, \quad x_i \geq 0 \end{array}} ?$$

[36.5] **Practice** Consider the problem  $\text{opt } \frac{x_1 + x_2^2 + x_3^3}{\text{s.t. } \begin{array}{l} x_1 + x_2 + x_3 = 1, \quad x_i \geq 0. \end{array}}$

a) Consider the minimization problem first and let  $a = (0, 1, 0)$ . Compute  $D(a)$ ,  $\mathcal{D}(a)$ ,  $Z(a)$ . Does KTCQ1 hold at  $a$ ? Is  $a$  a KT point? Is it a local minimum?

b) Can  $(0, 1, 0)$  be a point of local maximum?

c) Does KTCQ1 hold for every feasible point?

[36.6] **Practice** Consider the problem  $\max \frac{x^2 + y}{x \geq 0, x^2 + y^2 = 1}$ .

- Do you think, at each  $a \in T$ , KTCQ1 holds?
- Find the KT points.
- Conclude, whether these points are local maxima.

[36.7] **Practice** Find all local minimums for  $\min \frac{x^2 + y^2 + z^2 - x - y - z}{0 \leq x, y, z \leq 1}$ .

[36.8] **Exercise(E)** Consider the problems  $\min \frac{x^4}{|x| \leq 1}$  and  $\min \frac{x^3}{|x| \leq 1}$ .

- Argue that  $a = 0$  is a KT point for both.
- Now show that ktsc part a) is not a necessary condition. Hence, conclude that if  $d^t H(L)(a)d = 0$  for some nonzero  $d \in \mathcal{D}_+(a)$ , then  $a$  may or may not be a point of local minimum.
- However, show that we can still apply the ktsc part c) to conclude that 0 is a point of local minimum for the first One.

[36.9] **Exercise(M)** Consider the problem  $\max \frac{x_1 x_2 x_3 x_4}{x_1 + x_2 + x_3 + x_4 = 5, x_i \geq 0}$ . We already know one way to find the absolute maximum. Find all KT points and give another solution using KT theory. Also search for all local maximums.

[36.10] **Practice** Consider minimizing  $x^2 + y^2 + xy$  over the region  $T = \{x^2 + y^2 \geq 1, x^2 + 4y^2 \leq 4\}$ . Show that a point of local minimum must be a KT point.

[36.11] **Practice** Consider minimizing  $x^2 + y^2 + xy$  over the region  $T = \{x^2 + y^2 \geq 1, x^2 + y^2 \leq 4\}$ . Assume that the points of local minimums are KT points. Find all KT points. Apply KTSC at these points.

$$\exists w_j \in \mathbb{R} \text{ s.t. } \nabla f = \sum w_j \nabla h_j, \quad a \in T$$

Lagrange Method

$$\begin{bmatrix} \nabla h_1 & \nabla h_2 & \dots & \nabla h_p \end{bmatrix} \begin{bmatrix} w_1 \\ w_2 \\ \vdots \\ w_p \end{bmatrix} = \begin{bmatrix} \nabla f \end{bmatrix}$$

$$\downarrow \quad \underline{J^t w = \nabla f}$$

Consider the problem with only equality constraints.

$$(P3) \quad \begin{array}{ll} \text{opt} & f(x) \\ \text{s.t.} & h_j(x) = 0, j = 1, \dots, p \end{array}$$

The Lagrangian function here becomes  $L(x, w) = f(x) - \sum w_j h_j(x)$ . The matrix  $J := \begin{bmatrix} h'_1(a) \\ \vdots \\ h'_p(a) \end{bmatrix}$  is called the JACOBIAN.

[36.12] **Lemma** (Lagrangian necessary condition (LNC)) Let  $f, h_j \in \mathcal{C}^1$  and  $a$  be a local optimum for

213

It rank  $J = p < n$ .

LNC  $a$

$\{ \nabla g_i(a), \nabla h_j(a) \}$  active  $\rightarrow$  lin ind  $\Rightarrow$  KT(a) holds at  $a$ .

KTNLC  $\Rightarrow$  if  $KT(a)$  holds at  $a$ , and  $a$  is a local min, then  $a$  must be KT pt.

(P3). If the Jacobian  $J$  has full rank, then  $\exists w$  such that  $\nabla L(a, w) = 0$ , that is,  $a$  is a KT point.

*Proof.* Suppose that  $J$  has full rank. If  $p \leq n$ , then rows of  $J$  are linearly independent. So by licq, ktcq1 holds at  $a$ . Thus  $a$  being a local minimum must be a kt point.

Now suppose that  $p > n$ . Assume without loss that the first  $n$  rows of  $J$  are linearly independent. Then  $f'(a)$  being a vector in  $\mathbb{R}^n$ , is a combination  $\sum_{j=1}^n w_j h'_j(a)$ . Put  $w_j = 0$ , for  $j = n+1, \dots, p$ . ■

[36.13] **Theorem** (Lagrangian sufficient condition (LSC1)) Let  $f, h_j \in \mathcal{C}^2$  and suppose that  $a \in T$  satisfies  $\nabla L(a, w) = 0$ . Then

a)  $a$  is a strict local minimum if for each  $d \neq 0$  in  $\{d \mid Jd = 0\}$  we have  $d^t H_L(a) d > 0$ .

b)  $a$  is a strict local maximum if for each  $d \neq 0$  in  $\{d \mid Jd = 0\}$  we have  $d^t H_L(a) d < 0$ .

*Proof.* Follows from LSC, as  $\mathcal{D}_+(a) = \mathcal{D}(a) = \{d \mid Jd = 0\}$ . ■

KTSC a)

[36.14] **Discussion** a) We need to check whether  $d^t H_L d > 0$ , for any  $d \neq 0$  satisfying  $Jd = 0$ .

b) See, if  $\text{rank } J = n$ , then the above will not help, as  $d = 0$  is the only vector which will satisfy  $Jd = 0$ .

c) So, assume  $\text{rank } J < n$ . Let  $\hat{J}$  be the submatrix obtained from  $J$  by keeping a maximal linearly independent set of rows. Then  $\{d \mid \hat{J}d = 0\} = \{d \mid Jd = 0\}$ .

d) Under the assumption that  $\text{rank}(J) = p < n$ , using the symmetric nature of the matrix  $H_L(a)$ , a test was supplied by H. B. Mann, in 'Quadratic Forms with Linear Constraints', American Mathematical. Monthly (1943), pages 430–433. Accordingly, we have a sufficient condition for an optimal point.

[36.15] **Theorem** (Lagrangian sufficient condition 2 (LSC2)) Let  $f, h_j \in \mathcal{C}^2$ ,  $p < n$ , and  $a \in T$  satisfy  $\nabla L(a, w) = 0$ . Consider the BORDERED HESSIAN matrix  $B_{n+p \times n+p} = \begin{bmatrix} 0 & J \\ J^t & H_L \end{bmatrix}$ . Check the last  $n-p$  leading principal minors starting with the determinant of  $B$ .

a) If they have alternate signs starting with  $(-1)^n$ , then  $a$  is a strict local maximum.

b) If all of them have the sign  $(-1)^p$  then  $a$  is a strict local minimum.

$$B = \begin{bmatrix} 0 & J \\ J^t & H_L \end{bmatrix}$$

[36.16] **Example** Solve  $\min f = x + y$   
s.t.  $x^2 + y^2 = 1$ .

•  $J = \begin{bmatrix} 1 & 1 \end{bmatrix} \rightarrow \text{rank } 1 \text{ over } T$ .

LNC satisfied. so <sup>214</sup> a pt of opt min must be a KT pt.

$$-\begin{bmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{bmatrix} \text{ with } w = -\frac{1}{\sqrt{2}} \quad \Rightarrow \quad \lambda = \frac{1}{2w} = y, \quad x^2 + y^2 = 1$$

$$\checkmark \begin{bmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{bmatrix} \text{ with } w = \frac{1}{\sqrt{2}} \quad \Rightarrow \quad w = \frac{1}{\sqrt{2}}, \quad \lambda = \frac{1}{2w} = \frac{\sqrt{2}}{2} = \frac{1}{\sqrt{2}}, \quad \frac{1}{4w^2} \rightarrow \frac{1}{4w^2} = 1$$

$$xw^2 = 1, \quad w = \pm \frac{1}{\sqrt{2}}$$

Answer. Note that  $f, g \in \mathcal{C}^2$ .

◦ Matrix  $J = \begin{bmatrix} 2x & 2y \end{bmatrix}$  has full rank throughout the feasible set. So a local minimum must be a KT point.

◦ Find KT points:  $\begin{bmatrix} 1 \\ 1 \end{bmatrix} = w \begin{bmatrix} 2x \\ 2y \end{bmatrix} \Rightarrow x = y = \frac{1}{2w}$ . Since  $x^2 + y^2 = 1$ , we have  $w = \pm \frac{1}{\sqrt{2}}$ .

◦ Two possibilities:  $w = \frac{-1}{\sqrt{2}}, x = y = \frac{-1}{\sqrt{2}}$ , and  $w = \frac{1}{\sqrt{2}}, x = y = \frac{1}{\sqrt{2}}$ .

◦ The bordered Hessian matrix is  $B = \begin{bmatrix} 0 & 2x & 2y \\ 2x & -2w & 0 \\ 2y & 0 & -2w \end{bmatrix}$ .

◦ At  $w = \frac{1}{\sqrt{2}}, x = y = \frac{1}{\sqrt{2}}$ , the matrix is  $\begin{bmatrix} 0 & \sqrt{2} & \sqrt{2} \\ \sqrt{2} & -\sqrt{2} & 0 \\ \sqrt{2} & 0 & -\sqrt{2} \end{bmatrix}$ . As  $n - p = 1$ , we only check one leading minor starting from  $\det B$ . Now,  $\det B = 4\sqrt{2}$  has the sign  $(-1)^n$ . Hence  $x = y = \frac{1}{\sqrt{2}}$  is a point of maximum.

◦ Similarly, the other point can be shown to be a point of minimum.

◦ Note that if  $H_L$  is pd, we can directly conclude that a KT point is a point of strict local minimum. You do not need to use the bordered Hessian.

[36.17] **Example (Self)** Solve  $\min \frac{f = x + y + z}{x^2 + y^2 + z^2 = 1}$ .

Answer. Note that  $f, g \in \mathcal{C}^2$ .

◦ Matrix  $J = \begin{bmatrix} 2x & 2y & 2z \end{bmatrix}$  has full rank throughout the feasible set. So a local minimum is a KT point.

◦ To find KT points:  $\begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = w \begin{bmatrix} 2x \\ 2y \\ 2z \end{bmatrix} \Rightarrow x = y = z = \frac{1}{2w}$ . Since  $x^2 + y^2 + z^2 = 1$ , we have  $w = \pm \frac{\sqrt{3}}{2}$ .

◦ Two possibilities:  $w = \frac{-\sqrt{3}}{2}, x = y = z = \frac{-1}{\sqrt{3}}$ , and  $w = \frac{\sqrt{3}}{2}, x = y = z = \frac{1}{\sqrt{3}}$ .

◦ The bordered Hessian matrix is  $B = \begin{bmatrix} 0 & 2x & 2y & 2z \\ 2x & -2w & 0 & 0 \\ 2y & 0 & -2w & 0 \\ 2z & 0 & 0 & -2w \end{bmatrix}$ .

◦ At  $w = \frac{\sqrt{3}}{2}, x = y = z = \frac{1}{\sqrt{3}}$  the matrix is  $B = \begin{bmatrix} 0 & \frac{2}{\sqrt{3}} & \frac{2}{\sqrt{3}} & \frac{2}{\sqrt{3}} \\ \frac{2}{\sqrt{3}} & -\sqrt{3} & 0 & 0 \\ \frac{2}{\sqrt{3}} & 0 & -\sqrt{3} & 0 \\ \frac{2}{\sqrt{3}} & 0 & 0 & -\sqrt{3} \end{bmatrix}$ .

As  $n - p = 2$ , we find two leading minors starting from  $\det B$ , that is  $\det B$  and the determinant of the upper-left  $3 \times 3$  matrix  $C$ .

Now,  $\det B = -\sqrt{3} \det C - \frac{2}{\sqrt{3}} 2\sqrt{3} = -12$ ,  $\det C = \frac{8}{\sqrt{3}}$ . The signs alter starting from the sign  $(-1)^n$ . Hence this is a point of local maximum.

◦ Similarly, the other point can be shown to be a point of local minimum.

*Alternate.* Note that, at  $w = \frac{-\sqrt{3}}{2}$ ,  $H_L$  is pd. So by KTSC, we can directly conclude that this point is a point of strict minimum. Then use continuity over a compact set to conclude that the other point must be the absolute maximum.

## Exercises

[36.18] **Practice** Use Lagrange multipliers to solve

$$\begin{array}{ll} \text{opt} & f(x) \equiv 2x_1^2 + x_2^2 + 3x_3^2 + 10x_1 + 8x_2 + 6x_3 - 100 \\ \text{s.t.} & h \equiv x_1 + x_2 + x_3 - 20 = 0. \end{array}$$

[36.19] **Practice** Use Lagrange multipliers to solve  $\text{opt } f \equiv xy$   $\frac{\quad}{\text{s.t. } h \equiv x^2 + 4y^2 - 1 = 0}$ .

[36.20] **Practice** Maximize  $f(x, y, z) = x^2 + y^2 + z^2$ , subject to  $h_1 \equiv \frac{x^2}{4} + \frac{y^2}{5} + \frac{z^2}{25} - 1 = 0$ , and  $h_2 \equiv x + y - z = 0$ .

[36.21] **Practice** Consider the problem  $\text{opt } \frac{x^2 + y}{x^2 + y^2 = 1}$ .

- a) Solve it using graphical method.
- b) Do you think each point of local optimum is a KT point? Why?
- c) Find all KT points.
- d) Conclude whether these points are local minimums or maximums.
- e) Which of these are global maximums or minimums?

[36.22] **Exercise**

A) Consider an  $n \times n$ ,  $n \geq 4$  matrix of the form  $\begin{bmatrix} 0 & 2 & 2 & 2 \\ 2 & -1 & 1 & 1 \\ 2 & 1 & -1 & 1 \\ 2 & 1 & 1 & -1 \end{bmatrix}$ . Show that its determinant is

$(-1)^{n-1} 2^n (n-1)$ . You can subtract a multiple of remaining rows from first row. Another way is to use a known result. When  $D$  is invertible, we have

$$\begin{aligned} \det \begin{bmatrix} A & B \\ C & D \end{bmatrix} &= \det \left( \begin{bmatrix} A & B \\ C & D \end{bmatrix} \begin{bmatrix} I & 0 \\ -D^{-1}C & I \end{bmatrix} \right) = \det \begin{bmatrix} A - BD^{-1}C & B \\ 0 & D \end{bmatrix} \\ &\Rightarrow \det \begin{bmatrix} A & B \\ C & D \end{bmatrix} = \det D \det(A - BD^{-1}C) \end{aligned}$$

B) Solve  $\text{opt } \frac{x_1 x_2 x_3 x_4}{x_1^2 + x_2^2 + x_3^2 + x_4^2 = 4}$ .