

## 1 Recall

From Week 1, we discuss the follows:

**Euler's first order condition**

If  $f(\mathbf{x})$  is **continuously** differentiable,  $\emptyset \neq K$  is an open set in  $\mathbb{R}^n$  and  $\mathbf{x}^* \in K$  is an optimal solution to  $(P)$ , then

$$\nabla f(\mathbf{x}^*) = \mathbf{0}$$

In the previous lecture, we introduce a problem  $(P)$  and the *feasible* set  $K$  as follows:

$$\inf_{x \in \mathbb{R}^n} f(x) \quad \text{subject to} \quad \begin{cases} g_i(x) \leq 0, & i = 1, \dots, \ell \\ h_j(x) = 0, & j = 1, \dots, m \end{cases} \quad (P)$$

where  $f, g_i, h_j \in C^1$ , and

$$K = \{x \in \mathbb{R}^n : g_i(x) \leq 0, h_j(x) = 0, i = 1, \dots, \ell, j = 1, \dots, m\}$$

Also, we have the following theorems related to KKT condition and proved in the last lecture.

**Theorem 1.** Assume that  $x^* \in K$  is an optimal solution to  $(P)$ , then there exists  $p_0 \geq 0, p_1, \dots, p_\ell \geq 0, q_1, \dots, q_m \in \mathbb{R}$  such that the following holds:

1.  $(p_0, p_1, \dots, p_\ell, q_1, \dots, q_m) \neq \mathbf{0}$
2.  $\sum_{i=1}^{\ell} p_i g_i(x^*) = 0 \iff p_i g_i(x^*) = 0, \forall i = 1, 2, \dots, \ell$
3.  $p_0 \nabla f(x^*) + \sum_{i=1}^{\ell} p_i \nabla g_i(x^*) + \sum_{j=1}^m q_j \nabla h_j(x^*) = \mathbf{0}$

## 2 Qualification Condition

**Definition 1.** We say the constraints  $K$  is **qualified** at  $x \in K$  if  $p_i \geq 0$  and  $q_j \in \mathbb{R}$  satisfy

$$\begin{cases} \sum_{i=1}^{\ell} p_i g_i(x) = 0 \\ \sum_{i=1}^{\ell} p_i \nabla g_i(x) + \sum_{j=1}^m q_j \nabla h_j(x) = \mathbf{0} \end{cases}$$

then it implies that  $p_1 = \dots = p_\ell = q_1 = \dots = q_m = 0$ .

Now, when  $(P)$  has an optimal solution  $x^* \in K$  and constraints  $K$  is **qualified**, the following theorem is introduced.

**Theorem 2.** *Let  $x^* \in K$  be a solution to  $(P)$  and assume that  $K$  is **qualified** at  $x^*$ . Then there exists  $\lambda_1, \dots, \lambda_\ell \geq 0$  and  $\mu_1, \dots, \mu_m \in \mathbb{R}$  such that*

$$\begin{cases} \sum_{i=1}^{\ell} \lambda_i g_i(x^*) = 0 \\ \nabla f(x^*) + \sum_{i=1}^{\ell} \lambda_i \nabla g_i(x^*) + \sum_{j=1}^m \mu_j \nabla h_j(x^*) = \mathbf{0} \end{cases}$$

*Proof.* From Theorem 1, if  $x^* \in K$  is a solution to  $(P)$  and  $K$  is **qualified** at  $x^*$ , there exist

$$(p_0, p_1, \dots, p_\ell, q_1, \dots, q_m) \neq \mathbf{0}$$

such that

$$\begin{cases} \sum_{i=1}^{\ell} p_i g_i(x^*) = 0 \\ p_0 \nabla f(x^*) + \sum_{i=1}^{\ell} p_i \nabla g_i(x^*) + \sum_{j=1}^m q_j \nabla h_j(x^*) = \mathbf{0} \end{cases} \quad (*)$$

Now, if  $p_0 = 0$ , then  $(*)$  becomes

$$\begin{cases} \sum_{i=1}^{\ell} p_i g_i(x^*) = 0 \\ \sum_{i=1}^{\ell} p_i \nabla g_i(x^*) + \sum_{j=1}^m q_j \nabla h_j(x^*) = \mathbf{0} \end{cases}$$

By the Qualification condition, this implies that  $p_1 = \dots = p_\ell = q_1 = \dots = q_m = 0$  and hence

$$(p_0, p_1, \dots, p_\ell, q_1, \dots, q_m) = \mathbf{0}$$

Contradiction arises! Thus, we have  $p_0 > 0$ .

Dividing the second equality of  $(*)$  by  $p_0$  gives

$$\nabla f(x^*) + \sum_{i=1}^{\ell} \frac{p_i}{p_0} \nabla g_i(x^*) + \sum_{j=1}^m \frac{q_j}{p_0} \nabla h_j(x^*) = \mathbf{0}$$

Thus, it is natural to put  $\lambda_i = \frac{p_i}{p_0} \geq 0$  for  $i = 1, \dots, \ell$  and  $\mu_j = \frac{q_j}{p_0} \in \mathbb{R}$  for  $j = 1, \dots, m$ . □

Let's see a simple example for the importance of qualification condition.

**Example 1.** Solve the following problem

$$\min_{x \in \mathbb{R}} x, \quad \text{subject to} \quad x^2 = 0$$

**Solution.** As there is one variable problem, so putting  $n = 1$ ,  $f(x) = x$ . Since there is no inequality constraint, so  $\ell = 0$ . There is one equality constraint, so  $m = 1$ , and let  $h_1(x) = x^2$ .

It is easy to see that  $x^* = 0$  is an optimal solution to the problem as  $\{x \in \mathbb{R} : x^2 = 0\} = \{0\}$ .

From Theorem 1, there exists  $(p_0, q_1) \neq \mathbf{0}$  such that

$$p_0 \underbrace{\nabla f(x^*)}_{f'(0)=1} + q_1 \underbrace{\nabla h_1(x^*)}_{h'_1(0)=0} = \mathbf{0}$$

Thus,  $p_0 = 0$ , which means the qualification condition of the Theorem 2 is not satisfied. ◀

### 3 Formal Justification of Theorem 2

For the expansion of Theorem 2, it looks quite similar as “Lagrangian”. Indeed, we can rewrite the problem  $\min_{\substack{g(x) \leq 0 \\ h(x)=0}} f(x)$  as follows:

$$\begin{aligned} \min_{g(x) \leq 0, h(x)=0} f(x) &= \min_{x \in \mathbb{R}^k} \left( \sup_{\lambda \geq 0, \mu \in \mathbb{R}} f(x) + \lambda \cdot g(x) + \mu \cdot h(x) \right) \\ &\stackrel{(?)}{=} \sup_{\lambda \geq 0, \mu \in \mathbb{R}} \left( \min_{x \in \mathbb{R}} f(x) + \lambda \cdot g(x) + \mu \cdot h(x) \right) \\ &= \min_{x \in \mathbb{R}} f(x) + \lambda^* g(x) + \mu^* h(x) \end{aligned}$$

Denote  $L(x, \lambda, \mu) = f(x) + \lambda \cdot g(x) + \mu \cdot h(x)$ . If the feasible set  $K$  is qualified at  $x^*$ , then Theorem 2 implies  $\nabla L(x^*, \lambda^*, \mu^*) = \mathbf{0}$ , and such  $\lambda^*, \mu^*$  are called the **Language Multiplier**.

Now, let us do some exercises together on minimization problems subject to different constraints.

### 4 Exercises

**Exercise 1.** Solve the following problem

$$\min_{x^2+y^2=1} (2x + y)$$

**Solution.** As there are two variables, so putting  $n = 2$ , and let  $f(x) = 2x + y$ .

Moreover, there is no inequality constraint, so  $\ell = 0$ .

There is one equality constraint, so  $m = 1$  and let  $h(x, y) = x^2 + y^2 - 1$ .

If  $(x^*, y^*) \in K$  is an optimal solution (let us assume that  $K$  is qualified at  $(x^*, y^*)$  without checking), then by Theorem 2, there exists  $\mu_1 \in \mathbb{R}$  such that

$$\begin{aligned} \nabla f(x^*, y^*) + \mu_1 \nabla h(x^*, y^*) &= \mathbf{0} \\ \begin{pmatrix} 2 \\ 1 \end{pmatrix} + \mu_1 \begin{pmatrix} 2x^* \\ 2y^* \end{pmatrix} &= \begin{pmatrix} 0 \\ 0 \end{pmatrix} \end{aligned}$$

Therefore, we have

$$\begin{cases} 1 + \mu_1 x^* = 0 \\ 1 + 2\mu_1 y^* = 0 \\ (x^*)^2 + (y^*)^2 = 1 \end{cases}$$

Note that  $\mu_1 \neq 0$ , and from the first two equations, we have

$$x^* = -\frac{1}{\mu_1} \quad \text{and} \quad y^* = -\frac{1}{2\mu_1}$$

Putting into the third equation yields:

$$\begin{aligned} \left(-\frac{1}{\mu_1}\right)^2 + \left(-\frac{1}{2\mu_1}\right)^2 &= 1 \\ \mu_1^2 &= \frac{5}{4} \\ \mu_1 &= \pm \frac{\sqrt{5}}{2} \end{aligned}$$

So, we have  $(x^*, y^*) = \left(\frac{2}{\sqrt{5}}, \frac{1}{\sqrt{5}}\right)$  or  $\left(-\frac{2}{\sqrt{5}}, -\frac{1}{\sqrt{5}}\right)$ .

It remains to compare the values of  $f\left(\frac{2}{\sqrt{5}}, \frac{1}{\sqrt{5}}\right)$  and  $f\left(-\frac{2}{\sqrt{5}}, -\frac{1}{\sqrt{5}}\right)$  to get the minimum.

By direct computation, we find that  $(x^*, y^*) = \left(-\frac{2}{\sqrt{5}}, -\frac{1}{\sqrt{5}}\right)$  and  $f(x^*, y^*) = -\sqrt{5}$ . ◀

**Exercise 2.** Solve the problem  $\min_{x^2+y^2 \leq 1} x \cdot y$ .

**Solution.** Letting  $f(x, y) = xy$ ,  $\ell = 1$ ,  $m = 0$  and  $g(x) = x^2 + y^2 - 1$ .

Assume the feasible set  $K := \{(x, y) \in \mathbb{R}^2 : x^2 + y^2 \leq 1\}$  is *qualified*, then by Theorem 2, there exists  $\lambda \geq 0$  such that

$$\begin{cases} \lambda \cdot g(x, y) = 0 \\ \nabla f(x, y) + \lambda g(x, y) = \mathbf{0} \end{cases} \implies \begin{cases} \lambda(x^2 + y^2 - 1) = 0 \\ \begin{pmatrix} y \\ x \end{pmatrix} + \lambda \begin{pmatrix} 2x \\ 2y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \end{cases}$$

Now, we separate it into different cases:

- **Case 1:**  $\lambda = 0$

In the second equation, this leads to  $(x, y) = (0, 0)$ .

- **Case 2:**  $\lambda > 0$

Then, we have

$$\begin{cases} x^2 + y^2 = 1 \\ y + 2\lambda x = 0 \\ x + 2\lambda y = 0 \end{cases}$$

Combining the second and the third equations gives  $y + 2\lambda(-2\lambda y) = 0 \implies (1 - 4\lambda^2)y = 0$ .

Hence, we have  $\lambda = \frac{1}{2}$  and  $y \neq 0$  (*think why?*). As  $\lambda = \frac{1}{2}$ , this would follow that  $x = -y$ .

Putting back to  $x^2 + y^2 = 1$ , we have  $x = \pm \frac{1}{\sqrt{2}}$  and so  $y = \mp \frac{1}{\sqrt{2}}$ .

So, we now have  $(x, y) = (0, 0), (1/\sqrt{2}, -1/\sqrt{2})$  and  $(-1/\sqrt{2}, 1/\sqrt{2})$ .

By comparing values on those points to evaluate  $x \cdot y$ , both  $(1/\sqrt{2}, -1/\sqrt{2})$  and  $(-1/\sqrt{2}, 1/\sqrt{2})$  are the optimal solutions, and  $\min_{x^2+y^2 \leq 1} x \cdot y = -\frac{1}{2}$ . ◀

**Exercise 3.** Solve the following problem

$$\min_{\substack{x^2+y^2=1 \\ y^2+z^2=4}} (x+z)$$

**Solution.** As there are three variables, so putting  $n = 3$ , and  $f(x, y, z) = x + z$ . Since there is no inequality constraint but 2 equality constraints, so we have  $\ell = 0$ ,  $m = 2$  and letting

$$\begin{aligned} h_1(x, y, z) &= x^2 + y^2 - 1 \\ h_2(x, y, z) &= y^2 + z^2 - 4 \end{aligned}$$

By Theorem 2 (without checking qualification condition), there exists  $\mu_1, \mu_2 \in \mathbb{R}$  such that

$$\begin{aligned} &\begin{cases} \nabla f(x, y, z) + \mu_1 \nabla h_1(x, y, z) + \mu_2 \nabla h_2(x, y, z) = \mathbf{0} \\ x^2 + y^2 = 1 \\ y^2 + z^2 = 4 \end{cases} \\ \implies &\begin{cases} \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} + \mu_1 \begin{pmatrix} 2x \\ 2y \\ 0 \end{pmatrix} + \mu_2 \begin{pmatrix} 0 \\ 2y \\ 2z \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \\ x^2 + y^2 = 1 \\ y^2 + z^2 = 4 \end{cases} \\ \implies &\begin{cases} 1 + 2\mu_1 x = 0 \\ 2(\mu_1 + \mu_2)y = 0 \\ 1 + 2\mu_2 z = 0 \\ x^2 + y^2 = 1 \\ y^2 + z^2 = 4 \end{cases} \end{aligned}$$

Now, we consider into several cases:

- **Case 1:**  $y = 0$

Then, we have  $x = \pm 1$  and  $z = \pm 2$ . So, the solution are  $(x, y, z) = (\pm 1, 0, \pm 2)$ .

- **Case 2:**  $y \neq 0$

Then, we have  $\mu_1 = -\mu_2$ . Plug into the first and the third equations, we have  $x = -z = -\frac{1}{2\mu_1}$ .

In this case, there exists  $\mu_1, \mu_2 \in \mathbb{R} \setminus \{0\}$  such that  $x + z \equiv 0$ .

By comparing  $f(\pm 1, 0, \pm 2)$  and 0, simple calculation gives the optimal solution  $(x^*, y^*, z^*) = (-1, 0, -2)$  and  $\min_{\substack{x^2+y^2=1 \\ y^2+z^2=4}} (x+z) = -3$ . ◀

— End of Lecture 3 —