

# 最优化的一种新途径

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**摘要** 本文用 $n$ 维欧氏空间 $R^n$ 中的隐函数定理研究等式约束问题的最优性必要条件,从而得出解这类问题的一种新途径。它较经典的Lagrange乘子法可减少解方程组的维数。

**关键词** 隐函数定理, 等式约束问题, 最优化, 欧几里得空间, Lagrange乘子

## 引言

本文在研究乘积Banach空间中的优化问题时,从抽象空间得出的结果推得关系式

$$P = (M^{-1}N)^T Q \quad (1.1)$$

现在直接用 $R^n$ 中的隐函数定理来导出这个结论。

## 1 最优性必要条件

我们知道,对方程组

$$h(x) = 0$$

有下面著名的隐函数定理,其中 $x \in R^n$ ,  $h: R^n \rightarrow R^m$ , 设 $m < n$ , 记 $h(x) = (h_1(x), \dots, h_m(x))$ 。

引理(隐函数定理)<sup>[2]</sup>。设 $x^0 = (x_1^0, \dots, x_n^0) \in R^n$  满足:

- (1) 在 $x^0$ 的某个邻域中 $h \in C^p$ ,  $p \geq 1$ ;
- (2)  $h(x^0) = 0$ ;
- (3)  $m \times m$ 阶 Jacobian 矩阵;

$$J = \begin{pmatrix} \frac{\partial h_1(x^0)}{\partial x_1} & \dots & \frac{\partial h_1(x^0)}{\partial x_m} \\ \vdots & & \vdots \\ \frac{\partial h_m(x^0)}{\partial x_1} & \dots & \frac{\partial h_m(x^0)}{\partial x_m} \end{pmatrix}$$

是非奇异的。

则存在  $\hat{x}^0 = (x_{m+1}^0, \dots, x_n^0) \in \mathbb{R}^{n-m}$  的一个邻域  $U$ , 使得  $x_1, \dots, x_m$  是  $\hat{x} = (x_{m+1}, \dots, x_n) \in U$  的函数, 记成  $x_i = \phi_i(\hat{x}), i = 1, \dots, m$ , 具有下列性质:

- (1)  $\phi_i \in C^p$ ;
- (2)  $x_i^0 = \phi_i(\hat{x}^0), i = 1, \dots, m$ ;
- (3)  $h_i(\phi_1(\hat{x}), \dots, \phi_m(\hat{x}), \hat{x}) = 0, i = 1, \dots, m, \hat{x} \in U$ .

现在考虑等式约束问题

$$(PE) \begin{cases} \min f(x) \\ s.t. g(x) = 0 \end{cases}$$

其中  $x \in \mathbb{R}^n, f: \mathbb{R}^n \rightarrow \mathbb{R}, g: \mathbb{R}^n \rightarrow \mathbb{R}^m, m < n$ . 记  $g(x) = (g_1(x), \dots, g_m(x)), p = n - m$ , 令

$$P = \left( \frac{\partial f(x^0)}{\partial x_1}, \dots, \frac{\partial f(x^0)}{\partial x_p} \right)^T,$$

$$Q = \left( \frac{\partial f(x^0)}{\partial x_{p+1}}, \dots, \frac{\partial f(x^0)}{\partial x_n} \right)^T,$$

$$N = \begin{pmatrix} \frac{\partial g_1(x^0)}{\partial x_1} & \dots & \frac{\partial g_1(x^0)}{\partial x_p} \\ \vdots & & \vdots \\ \frac{\partial g_m(x^0)}{\partial x_1} & \dots & \frac{\partial g_m(x^0)}{\partial x_p} \end{pmatrix}$$

$$M = \begin{pmatrix} \frac{\partial g_1(x^0)}{\partial x_{p+1}} & \dots & \frac{\partial g_1(x^0)}{\partial x_n} \\ \vdots & & \vdots \\ \frac{\partial g_m(x^0)}{\partial x_{p+1}} & \dots & \frac{\partial g_m(x^0)}{\partial x_n} \end{pmatrix}$$

定理 1 设  $x^0 \in \mathbb{R}^n$  是 (PE) 的最优解,  $f: \mathbb{R}^n \rightarrow \mathbb{R}, g: \mathbb{R}^n \rightarrow \mathbb{R}^m$  连续可微, 矩阵  $M$  非奇异, 则有

$$P = (M^{-1}N)^T Q$$

证 因  $g(x^0) = 0, M$  非奇异, 于是由隐函数定理  $\exists \hat{x}^0 = (x_{m+1}^0, \dots, x_n^0)$  的一个邻域  $U$ , 使得  $x_{p+1}, \dots, x_n$  是  $\hat{x} = (x_{m+1}, \dots, x_n) \in U$  的函数, 即有

$$x_i = \phi_i(\hat{x}), \hat{x} \in U, i = p+1, \dots, n$$

这些函数具有下列性质

- (1)  $\phi_i \in C^1$ ;
- (2)  $x_i^0 = \phi_i(\hat{x}^0), i = p+1, \dots, n$ ;
- (3)  $g_i(\phi_1(\hat{x}), \dots, \phi_n(\hat{x})) = 0, i = 1, \dots, m, \hat{x} \in U$ . (1)

因  $x^0$  是 (PE) 的最优解, 即

$$f(x^0) \leq f(x) \quad \forall x \in \{x \in \mathbb{R}^n : g(x) = 0\}$$

所以, 有

$$\begin{aligned} & f(\hat{x}^0, \varphi_{p+1}(\hat{x}^0), \dots, \varphi_n(\hat{x}^0)) \\ & \leq f(\hat{x}, \varphi_{p+1}(\hat{x}), \dots, \varphi_n(\hat{x})) \quad \forall x \in U \end{aligned} \quad (2)$$

令

$$F(\hat{x}) = f(\hat{x}, \varphi_{p+1}(\hat{x}), \dots, \varphi_n(\hat{x})), \quad \hat{x} \in U$$

由(2)式得

$$F(\hat{x}^0) \leq F(\hat{x}) \quad \forall \hat{x} \in U$$

故  $\hat{x}^0$  是  $F(\hat{x})$  的极小点, 于是,

$$\frac{\partial F(\hat{x}^0)}{\partial x_i} = 0, \quad i = 1, 2, \dots, p$$

由此得

$$\begin{pmatrix} \frac{\partial f(x^0)}{\partial x_1} \\ \vdots \\ \frac{\partial f(x^0)}{\partial x_p} \end{pmatrix} + \begin{pmatrix} \frac{\partial \varphi_{p+1}(\hat{x}^0)}{\partial x_1} \dots \frac{\partial \varphi_n(\hat{x}^0)}{\partial x_1} \\ \vdots \\ \frac{\partial \varphi_{p+1}(\hat{x}^0)}{\partial x_p} \dots \frac{\partial \varphi_n(\hat{x}^0)}{\partial x_p} \end{pmatrix} \begin{pmatrix} \frac{\partial f(x^0)}{\partial x_{p+1}} \\ \vdots \\ \frac{\partial f(x^0)}{\partial x_n} \end{pmatrix} = 0 \quad (3)$$

由(1)式同样有

$$\begin{pmatrix} \frac{\partial g_i(x^0)}{\partial x_1} \\ \vdots \\ \frac{\partial g_i(x^0)}{\partial x_p} \end{pmatrix} + \begin{pmatrix} \frac{\partial \varphi_{p+1}(\hat{x}^0)}{\partial x_1} \dots \frac{\partial \varphi_n(\hat{x}^0)}{\partial x_1} \\ \vdots \\ \frac{\partial \varphi_{p+1}(\hat{x}^0)}{\partial x_p} \dots \frac{\partial \varphi_n(\hat{x}^0)}{\partial x_p} \end{pmatrix} \begin{pmatrix} \frac{\partial g_i(x^0)}{\partial x_{p+1}} \\ \vdots \\ \frac{\partial g_i(x^0)}{\partial x_n} \end{pmatrix} = 0 \quad (4)$$

$i = 1, 2, \dots, m$

令

$$A = \begin{pmatrix} \frac{\partial \varphi_{p+1}(\hat{x}^0)}{\partial x_1} \dots \frac{\partial \varphi_n(\hat{x}^0)}{\partial x_1} \\ \vdots \\ \frac{\partial \varphi_{p+1}(\hat{x}^0)}{\partial x_p} \dots \frac{\partial \varphi_n(\hat{x}^0)}{\partial x_p} \end{pmatrix}$$

则(4)式变成

$$\begin{pmatrix} \frac{\partial g_i(x^0)}{\partial x_1} \\ \vdots \\ \frac{\partial g_i(x^0)}{\partial x_p} \end{pmatrix} = -A \begin{pmatrix} \frac{\partial g_i(x^0)}{\partial x_{p+1}} \\ \vdots \\ \frac{\partial g_i(x^0)}{\partial x_n} \end{pmatrix} \quad i = 1, \dots, m$$

从而

$$\begin{pmatrix} \frac{\partial g_1(x^0)}{\partial x_1} & \dots & \frac{\partial g_m(x^0)}{\partial x_1} \\ \vdots & & \vdots \\ \frac{\partial g_1(x^0)}{\partial x_p} & \dots & \frac{\partial g_m(x^0)}{\partial x_p} \end{pmatrix} = -A \begin{pmatrix} \frac{\partial g_1(x^0)}{\partial x_{p+1}} & \dots & \frac{\partial g_m(x^0)}{\partial x_{p+1}} \\ \vdots & & \vdots \\ \frac{\partial g_1(x^0)}{\partial x_n} & \dots & \frac{\partial g_m(x^0)}{\partial x_n} \end{pmatrix}$$

即

$$N^T = -AM^T$$

因 $M$ 非奇异, 所以

$$A = -N^T(M^T)^{-1} \quad (5)$$

最后, 根据(3)、(5)两式便得出

$$P = (M^{-1}N)^T Q \quad \text{证毕}$$

例 1

$$\begin{cases} \min f = x_1^2 + x_2^2 + x_3^2 \\ \text{s.t. } g_1 = x_1^2 + x_2^2 - x_3 = 0 \\ g_2 = x_1 + x_2 + x_3 - 1 = 0 \end{cases}$$

最优解是  $\left(\frac{-1+\sqrt{3}}{2}, \frac{-1+\sqrt{3}}{2}, 2-\sqrt{3}\right)$ . 这时,  $p = 3 - 2 = 1$ ,

$$M = \begin{bmatrix} -1+\sqrt{3} & -1 \\ 1 & 1 \end{bmatrix}, \quad M^{-1} = \frac{1}{\sqrt{3}} \begin{bmatrix} 1 & 1 \\ -1 & -1+\sqrt{3} \end{bmatrix}$$

$$N = (-1 + \sqrt{3}, 1)^T$$

$$P = -1 + \sqrt{3}$$

$$Q = (-1 + \sqrt{3}, 2(2 - \sqrt{3}))^T$$

从而

$$M^{-1}N = (1, 0)^T$$

$$(M^{-1}N)^T Q = -1 + \sqrt{3}$$

因此有  $P = (M^{-1}N)^T Q$ .

注 从定理 1 的证明看出, 矩阵  $P, Q, M, N$  中变量  $x_1, \dots, x_n$  的下标不一定要依次取, 关键在于约束函数  $g$  在点  $x^0$  的 Frechet 一导数要有一个  $m$  阶子块非奇异, 这等价于梯度向量组  $\nabla g_1(x^0), \dots, \nabla g_m(x^0)$  线性无关。

下面, 讨论等式  $P = (M^{-1}N)^T Q$  与 Kuhn-Tucker 条件的关系。

$P = N^T(M^{-1})^T Q$  即是

$$\begin{pmatrix} \frac{\partial f(x^0)}{\partial x_1} \\ \vdots \\ \frac{\partial f(x^0)}{\partial x_p} \end{pmatrix} = \begin{pmatrix} \frac{\partial g_1(x^0)}{\partial x_1} & \dots & \frac{\partial g_m(x^0)}{\partial x_1} \\ \vdots & & \vdots \\ \frac{\partial g_1(x^0)}{\partial x_p} & \dots & \frac{\partial g_m(x^0)}{\partial x_p} \end{pmatrix} (M^{-1})^T Q \quad (6)$$

然而

$$\begin{pmatrix} \frac{\partial g_1(x^0)}{\partial x_{p+1}} & \dots & \frac{\partial g_m(x^0)}{\partial x_{p+1}} \\ \vdots & & \vdots \\ \frac{\partial g_1(x^0)}{\partial x_n} & \dots & \frac{\partial g_m(x^0)}{\partial x_n} \end{pmatrix} (M^{-1})^T Q = M^T (M^{-1})^T Q = Q \quad (7)$$

所以由(6), (7)式有

$$\begin{pmatrix} \frac{\partial f(x^0)}{\partial x_1} \\ \vdots \\ \frac{\partial f(x^0)}{\partial x_n} \end{pmatrix} = \begin{pmatrix} \frac{\partial g_1(x^0)}{\partial x_1} & \dots & \frac{\partial g_m(x^0)}{\partial x_1} \\ \vdots & & \vdots \\ \frac{\partial g_1(x^0)}{\partial x_n} & \dots & \frac{\partial g_m(x^0)}{\partial x_n} \end{pmatrix} (M^{-1})^T Q \quad (8)$$

即是

$$\nabla f(x^0) = \nabla g(x^0)^T (M^{-1})^T Q$$

令  $\lambda = (\lambda_1, \dots, \lambda_m)^T = -(M^{-1})^T Q$ , 则由(8)式得

$$\nabla f(x^0) + \sum_{i=1}^m \lambda_i \nabla g_i(x^0) = 0 \quad (9)$$

因此, 若  $P = (M^{-1}N)^T Q$ , 那末(9)式自然成立, 故有下面定理。

定理2 若  $P = (M^{-1}N)^T Q$ , 则

$$\nabla f(x^0) + \lambda^T \nabla g(x^0) = 0$$

其中  $\lambda = -(M^{-1})^T Q$ .

于是, 根据定理2, 可用求解方程组

$$\begin{cases} P = (M^{-1}N)^T Q \\ g(x) = 0 \end{cases} \quad (10)$$

来得出等式约束问题(PE)的K-T点, 其中

$$P = \left( \frac{\partial f(x)}{\partial x_1}, \dots, \frac{\partial f(x)}{\partial x_p} \right)^T \quad Q = \left( \frac{\partial f(x)}{\partial x_{p+1}}, \dots, \frac{\partial f(x)}{\partial x_n} \right)^T$$

$$N = \begin{pmatrix} \frac{\partial g_1(x)}{\partial x_1} & \dots & \frac{\partial g_1(x)}{\partial x_p} \\ \vdots & & \vdots \\ \frac{\partial g_m(x)}{\partial x_1} & \dots & \frac{\partial g_m(x)}{\partial x_p} \end{pmatrix}$$

$$M = \begin{pmatrix} \frac{\partial g_1(x)}{\partial x_{p+1}} & \dots & \frac{\partial g_1(x)}{\partial x_n} \\ \vdots & & \vdots \\ \frac{\partial g_m(x)}{\partial x_{p+1}} & \dots & \frac{\partial g_m(x)}{\partial x_n} \end{pmatrix}$$

使  $M$  非奇异的解便是(PE)的K-T点。

例2 
$$\begin{cases} \min f = x_1^2 + x_2^2 + x_3^2 \\ \text{s.t. } g = x_1 + x_2 + x_3 - 1 = 0 \end{cases}$$

这时  $p = 3 - 1 = 2$

$$P = (2x_1, 2x_2)^T$$

$$Q = 2x_3$$

$$N = (1, 1)$$

$$M = 1$$

由 (10) 式得

$$x_1 = x_3$$

$$x_2 = x_3$$

$$x_1 + x_2 + x_3 - 1 = 0$$

(11)

此方程组的解为  $(\frac{1}{3}, \frac{1}{3}, \frac{1}{3})^T$ 。这不仅是  $K-T$  点且易验证还是最优解。

现在用经典的 Lagrange 乘子法求解例 2。令

$$L = x_1^2 + x_2^2 + x_3^2 + \lambda(x_1 + x_2 + x_3 - 1)$$

解方程组

$$\frac{\partial L}{\partial x_1} = 2x_1 + \lambda = 0$$

$$\frac{\partial L}{\partial x_2} = 2x_2 + \lambda = 0$$

$$\frac{\partial L}{\partial x_3} = 2x_3 + \lambda = 0$$

$$\frac{\partial L}{\partial \lambda} = x_1 + x_2 + x_3 - 1 = 0$$

(12)

得最优解  $x^0 = (\frac{1}{3}, \frac{1}{3}, \frac{1}{3})^T$

从上述两种算法看出, 前者需解方程组(11)的维数为 3, 而后者需解方程组(12)的维数是 4。事实上, 一般地, 方程组(10)的维数  $= p + m = n - m + m = n$ 。对规划 (PE) 用经典 Lagrange 乘子法时, 由 Lagrange 函数

$$L(x, \lambda) = f(x) + \sum_{i=1}^m \lambda_i g_i(x)$$

对各变元的偏导数构成的方程组是

$$\frac{\partial L}{\partial x_i} = 0, \quad i = 1, 2, \dots, n,$$

$$\frac{\partial L}{\partial \lambda_j} = 0, \quad j = 1, 2, \dots, m$$

其维数  $= n + m$ 。可见, 当  $m$  较大时, 两种方法所解相应方程组的维数差别是很大的, 从这个意义讲前者较后者简单。

$$\text{例 3} \quad \begin{cases} \min & f = x_1^2 + x_2^2 + x_3^2 \\ \text{s.t.} & g_1 = x_1^2 + x_2^2 - x_3 = 0 \\ & g_2 = x_1 + x_2 + x_3 - 1 = 0 \end{cases}$$

这时  $p = 3 - 2 = 1$ .

$$P = 2x_1$$

$$Q = (2x_2, 2x_3)^T$$

$$N = (2x_1, 1)^T$$

$$M = \begin{bmatrix} 2x_2 & -1 \\ 1 & 1 \end{bmatrix}$$

$M$  的行列式

$$|M| = 2x_2 + 1 \quad \left(x_2 \neq -\frac{1}{2}\right)$$

$M$  的逆矩阵

$$M^{-1} = \frac{1}{2x_2 + 1} \begin{bmatrix} 1 & 1 \\ -1 & 2x_2 \end{bmatrix}$$

于是, 由(10)式得

$$x_1 = \frac{1}{2x_2 + 1} (x_2(2x_1 + 1) - 2x_3(x_1 - x_2))$$

$$x_1^2 + x_2^2 - x_3 = 0$$

$$x_1 + x_2 + x_3 - 1 = 0$$

其实数解为  $\left(-\frac{1+\sqrt{3}}{2}, -\frac{1+\sqrt{3}}{2}, 2-\sqrt{3}\right)^T, \left(-\frac{1-\sqrt{3}}{2}, -\frac{1-\sqrt{3}}{2}, 2+\sqrt{3}\right)^T$ .

二者均是  $K-T$  点, 但前者是最优解, 后者非也。

### 参 考 文 献

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## A NEW APPROACH OF OPTIMIZATION

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**ABSTRACT** In this paper, a necessary optimality condition of problems with equality constraints is investigated by the implicit function theorem in  $n$ -dimensional Euclidean space. A new approach solving these problems is obtained. The number of dimensions of corresponding system or equations is less than the classic Lagrangian multiplier method.

**KEY WORDS** optimality condition, equality constraint, implicit function, Euclidean space, Lagrangian multiplier