

# 知识点

## 前言

对于一元函数的极值问题相信大家都十分熟悉，但是对于多元函数的极值问题可能就会比较陌生。大家都学过淑芬怎么可能陌生呢

对于没有限制条件的多元函数来说，只需要对函数求导即可，但是若有了限制条件，即函数的值要在一定条件下才能取到，则需要用到拉格朗日乘子法。

## 前置知识

多元函数的求导

多元函数极值的概念

二次型矩阵

隐函数存在定理

## 引理

设函数  $f(\vec{x})$

$\{\varphi_1(\vec{x}), \varphi_2(\vec{x}), \dots, \varphi_m(\vec{x})\}$  在区域  $D \subset \mathbb{R}^n$  ( $m < n$ ) 内具有各个连续偏导数，再设

$\{\vec{x}_0\} = \{x_1^0, x_2^0, \dots, x_n^0\} \in D$  为  $f(\vec{x})$  在约束条件

$\begin{cases} \varphi_1(\vec{x}) = 0 \\ \varphi_2(\vec{x}) = 0 \\ \dots \\ \varphi_m(\vec{x}) = 0 \end{cases}$  下的极值点，并且  $\varphi'(x_0)$  的秩为  $m$  则存在

常数  $\{\lambda_1, \lambda_2, \dots, \lambda_3\} \in \mathbb{R}$  使得在  $\vec{x}_0$  处成立下述等式  $\frac{\partial f(\vec{x}_0)}{\partial x_i} + \sum_{j=1}^m \lambda_j \frac{\partial \varphi_j(\vec{x}_0)}{\partial x_i} = 0 \quad (i=1, 2, \dots, n)$

$\{\varphi_j(\vec{x}_0)\} = 0 \quad (j=1, 2, \dots, m)$

## 证明

由于  $\varphi'(\vec{x}_0)$  的秩为  $m$  我们不妨设行列

$\frac{\partial (\varphi_1, \varphi_2, \dots, \varphi_m)}{\partial (x_{n-m+1}, x_{n-m+2}, \dots, x_n)}$  在  $x_0$  处不为零。因此，在  $\vec{x}_0$  的某个邻域内唯一确定一组具有各个连续偏导数的隐函

数  $\begin{cases} x_{n-m+1} = g_1(x_1, x_2, \dots, x_{n-m}), \\ x_{n-m+2} = g_2(x_1, x_2, \dots, x_{n-m}), \\ \dots \\ x_n = g_m(x_1, x_2, \dots, x_{n-m}) \end{cases}$  满足

$x_j^0 = g_j(x_1^0, x_2^0, \dots, x_n^0) \quad (j=n-m+1, n-m+2, \dots, n)$  且有  $\varphi_k(x_1, \dots, x_{n-m}, g_1(x_1, x_2, \dots, x_{n-m}), \dots, g_m(x_1, x_2, \dots, x_{n-m})) = 0$  将隐函数组代入  $f(\vec{x}_0)$

得  $f(x_1, \dots, x_{n-m}, g_1(x_1, x_2, \dots, x_{n-m}), \dots, g_m(x_1, x_2, \dots, x_{n-m}))$  因此  $\vec{x}_0$  是条件极值点转化为  $(x_1^0, x_2^0, \dots, x_{n-m}^0)$  为上述函数的通常极值点。

令  $\vec{x}_0'$  则对  $i=1, 2, \dots, n-m$

有  $\frac{\partial f(\vec{x}_0)}{\partial x_i} + \frac{\partial f(\vec{x}_0)}{\partial x_{n-m+1}} \cdots \frac{\partial f(\vec{x}_0)}{\partial x_n} + \frac{\partial f(\vec{x}_0)}{\partial g_1} \frac{\partial g_1}{\partial x_i} + \dots + \frac{\partial f(\vec{x}_0)}{\partial g_m} \frac{\partial g_m}{\partial x_i} = 0$  令  $\vec{g}(\vec{x}) = (g_1(\vec{x}), g_2(\vec{x}), \dots, g_m(\vec{x}))$  其中  $\vec{x} = (x_1, x_2, \dots, x_{n-m})$  将上述  $n-m$  个等式写成向量形式，

有 $\left(\frac{\partial f(\vec{x}_0)}{\partial x_1}, \dots, \frac{\partial f(\vec{x}_0)}{\partial x_n}\right)$  由于 $\vec{g}(\vec{x}_0)' = -\begin{pmatrix} \frac{\partial f(\vec{x}_0)}{\partial x_1} & \dots & \frac{\partial f(\vec{x}_0)}{\partial x_n} \end{pmatrix}$  是一个 \$m\$ 维行向量，我们可以将其记为 $\lambda_1, \lambda_2, \dots, \lambda_m$ 。将 $\lambda_1, \lambda_2, \dots, \lambda_m$  代入之前的式子 得
 
$$\begin{aligned} & \left(\frac{\partial f(\vec{x}_0)}{\partial x_1}, \dots, \frac{\partial f(\vec{x}_0)}{\partial x_n}\right) \cdot \begin{pmatrix} \lambda_1 & \dots & \lambda_n \end{pmatrix} \\ &= \left(\frac{\partial f(\vec{x}_0)}{\partial x_1} \lambda_1 + \dots + \frac{\partial f(\vec{x}_0)}{\partial x_n} \lambda_n\right) \begin{pmatrix} 1 & \dots & 1 \end{pmatrix} \\ &= \left(\frac{\partial f(\vec{x}_0)}{\partial x_1} \lambda_1 + \dots + \frac{\partial f(\vec{x}_0)}{\partial x_n} \lambda_n\right) \cdot \frac{1}{n} \begin{pmatrix} n & \dots & n \end{pmatrix} \\ &= \frac{1}{n} \left( \lambda_1 \left(\frac{\partial f(\vec{x}_0)}{\partial x_1} + \dots + \frac{\partial f(\vec{x}_0)}{\partial x_n}\right) + \dots + \lambda_n \left(\frac{\partial f(\vec{x}_0)}{\partial x_1} + \dots + \frac{\partial f(\vec{x}_0)}{\partial x_n}\right) \right) \\ &= \frac{1}{n} \left( \lambda_1 \frac{\partial f(\vec{x}_0)}{\partial x_1} + \dots + \lambda_n \frac{\partial f(\vec{x}_0)}{\partial x_1} + \lambda_1 \frac{\partial f(\vec{x}_0)}{\partial x_2} + \dots + \lambda_n \frac{\partial f(\vec{x}_0)}{\partial x_2} + \dots + \lambda_1 \frac{\partial f(\vec{x}_0)}{\partial x_n} + \lambda_n \frac{\partial f(\vec{x}_0)}{\partial x_n} \right) \\ &= \frac{1}{n} \left( \lambda_1 \frac{\partial f(\vec{x}_0)}{\partial x_1} + \lambda_2 \frac{\partial f(\vec{x}_0)}{\partial x_2} + \dots + \lambda_n \frac{\partial f(\vec{x}_0)}{\partial x_n} \right) \\ &= \frac{1}{n} \left( \lambda_1 \frac{\partial f(\vec{x}_0)}{\partial x_1} + \lambda_2 \frac{\partial f(\vec{x}_0)}{\partial x_2} + \dots + \lambda_n \frac{\partial f(\vec{x}_0)}{\partial x_n} \right) \cdot \frac{1}{n} \begin{pmatrix} n & \dots & n \end{pmatrix} \\ &= \frac{1}{n^2} \left( \lambda_1 \frac{\partial f(\vec{x}_0)}{\partial x_1} + \lambda_2 \frac{\partial f(\vec{x}_0)}{\partial x_2} + \dots + \lambda_n \frac{\partial f(\vec{x}_0)}{\partial x_n} \right)^2 \end{aligned}$$

$\left(\begin{array}{l} \left.\left(4\right)\right| \\ \left.\left(5\right)\right| \end{array}\right)$  写成分量形式再加上约束条件即可证明。

## 拉格朗日乘子法

构造函数  $F(x_1, \dots, x_n, \lambda_1, \dots, \lambda_m) = f(\vec{x}) + \sum_{j=1}^m \lambda_j \varphi_j(\vec{x})$  则上述求条件极值点的必要条件形式转化为  $F$  的通常极值的必要条件  
$$\begin{cases} \frac{\partial F(\vec{x}_0)}{\partial x_i} = 0 & \text{for } i = 1, 2, \dots, n \\ \frac{\partial F(\vec{x}_0)}{\partial \lambda_j} = 0 & \text{for } j = 1, 2, \dots, m \end{cases}$$
 此即拉格朗日乘子法

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