

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。

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 $\begin{cases} \varphi_1(\vec{x}) = 0 \\ \varphi_2(\vec{x}) = 0 \\ \vdots \\ \varphi_m(\vec{x}) = 0 \end{cases}$ 下的极值点，并且 $\varphi_i'(x_0)$ 的秩为 m ，则存在常数 $\lambda_1, \lambda_2, \dots, \lambda_m$ 使得在 \vec{x}_0 处成立下述等式 $\begin{cases} \frac{\partial f(\vec{x}_0)}{\partial x_i} + \sum_{j=1}^m \lambda_j \frac{\partial \varphi_j(\vec{x}_0)}{\partial x_i} = 0 \quad (i=1, 2, \dots, n) \\ \varphi_j(\vec{x}_0) = 0 \quad (j=1, 2, \dots, m) \end{cases}$

证明

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

$\begin{cases} \frac{\partial (\varphi_1, \varphi_2, \dots, \varphi_m)}{\partial (x_{n-m+1}, x_{n-m+2}, \dots, x_n)} \neq 0 \\ \vec{x}_0 \text{ 处不为零} \end{cases}$ 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 $\begin{cases} x_j^0 = g_j(x_1^0, x_2^0, \dots, x_{n-m}^0) \quad (j=n-m+1, n-m+2, \dots, n) \\ g_1(x_1^0, x_2^0, \dots, x_{n-m}^0), \dots, g_m(x_1^0, x_2^0, \dots, x_{n-m}^0) = 0 \end{cases}$ 将隐函数组代入 $f(\vec{x}_0)$ 得 $f(x_1^0, \dots, x_{n-m}^0, g_1(x_1^0, x_2^0, \dots, x_{n-m}^0), \dots, g_m(x_1^0, x_2^0, \dots, x_{n-m}^0))$ ，因此 \vec{x}_0 是条件极值点转化为 $(x_1^0, x_2^0, \dots, x_{n-m}^0)^0$ 为上述函数的通常极值点。

令 \vec{x}_0' 则对 $i=1, 2, \dots, n-m$

$\begin{aligned} & \frac{\partial f(\vec{x}_0)}{\partial x_i} + \frac{\partial f(\vec{x}_0)}{\partial x_{n-m+1}} \frac{\partial g_1}{\partial x_i} + \dots + \frac{\partial f(\vec{x}_0)}{\partial x_n} \frac{\partial g_m}{\partial x_i} = 0 \\ & \vec{g}' = (g_1', g_2', \dots, g_m') = (g_1(\vec{x}_0'), g_2(\vec{x}_0'), \dots, g_m(\vec{x}_0'))^T \quad \text{其中} \\ & \vec{x}' = (x_1, x_2, \dots, x_{n-m}) \quad \text{将上述 } n-m \text{ 个等式写成向量形式,} \\ & \left(\frac{\partial f(\vec{x}_0)}{\partial x_1}, \frac{\partial f(\vec{x}_0)}{\partial x_2}, \dots, \frac{\partial f(\vec{x}_0)}{\partial x_{n-m}} \right)' = \vec{g}' \\ & \left(\frac{\partial f(\vec{x}_0)}{\partial x_1}, \frac{\partial f(\vec{x}_0)}{\partial x_2}, \dots, \frac{\partial f(\vec{x}_0)}{\partial x_{n-m}} \right)' \vec{g}' = 0 \quad \text{由于 } \vec{g}' = -\left(\frac{\partial f(\vec{x}_0)}{\partial x_{n-m+1}}, \dots, \frac{\partial f(\vec{x}_0)}{\partial x_n} \right)' \end{aligned}$

\frac{\partial{\varphi_2(\vec{x_0})}}{\partial{x_{n-m+2}}} \cdots &
\frac{\partial{\varphi_2(\vec{x_0})}}{\partial{x_n}} \vdots \vdots \ddots \vdots \\ \frac{\partial{\varphi_m(\vec{x_0})}}{\partial{x_{n-m+1}}} &
\frac{\partial{\varphi_m(\vec{x_0})}}{\partial{x_{n-m+2}}} \cdots &\cdots &
\frac{\partial{\varphi_m(\vec{x_0})}}{\partial{x_n}}\end{array}\right)^{-1} \left.\begin{array}{l} \frac{\partial{\varphi_1(\vec{x_0})}}{\partial{x_1}} &
\frac{\partial{\varphi_1(\vec{x_0})}}{\partial{x_2}} \cdots &\cdots &
\frac{\partial{\varphi_1(\vec{x_0})}}{\partial{x_{n-m}}} \\\frac{\partial{\varphi_2(\vec{x_0})}}{\partial{x_1}} &
\frac{\partial{\varphi_2(\vec{x_0})}}{\partial{x_2}} \cdots &\cdots &
\frac{\partial{\varphi_2(\vec{x_0})}}{\partial{x_{n-m}}} \\\vdots &\vdots &\vdots &\vdots \\ \frac{\partial{\varphi_m(\vec{x_0})}}{\partial{x_1}} &
\frac{\partial{\varphi_m(\vec{x_0})}}{\partial{x_2}} \cdots &\cdots &
\frac{\partial{\varphi_m(\vec{x_0})}}{\partial{x_{n-m}}}\end{array}\right)\triangleq -A^{-1}B\quad \text{注意到 } A^{-1}B \text{ 是一个 } m \text{ 维行向量，我们可以将其记为 } A^{-1}B = \left(\frac{\partial f(\vec{x_0})}{\partial x_1}, \dots, \frac{\partial f(\vec{x_0})}{\partial x_n}\right)^T. \text{ 将 } A^{-1}B \text{ 代入之前的式子 } A^{-1}B^T A \text{ 得 }
A^{-1}B^T A = \left(\frac{\partial f(\vec{x_0})}{\partial x_1}, \dots, \frac{\partial f(\vec{x_0})}{\partial x_n}\right)^T \left(\frac{\partial f(\vec{x_0})}{\partial x_1}, \dots, \frac{\partial f(\vec{x_0})}{\partial x_n}\right) = \sum_{i=1}^n \left(\frac{\partial f(\vec{x_0})}{\partial x_i}\right)^2 = \|f(\vec{x_0})\|^2. \text{ 另外我们可以将 } A^{-1}B^T A \text{ 改写成 } A^{-1}B^T A = \left(\frac{\partial f(\vec{x_0})}{\partial x_1}, \dots, \frac{\partial f(\vec{x_0})}{\partial x_n}\right)^T \left(\frac{\partial f(\vec{x_0})}{\partial x_1}, \dots, \frac{\partial f(\vec{x_0})}{\partial x_n}\right) = \sum_{i=1}^n \left(\frac{\partial f(\vec{x_0})}{\partial x_i}\right)^2 = \|f(\vec{x_0})\|^2. \text{ 将 } A^{-1}B^T A \text{ 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 \quad (i=1,2,\dots,n) \\ \frac{\partial F(\vec{x}_0)}{\partial \lambda_j} = 0 \quad (j=1,2,\dots,m) \end{cases}\$\$ 此即拉格朗日乘子法

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