

最优化理论与方法

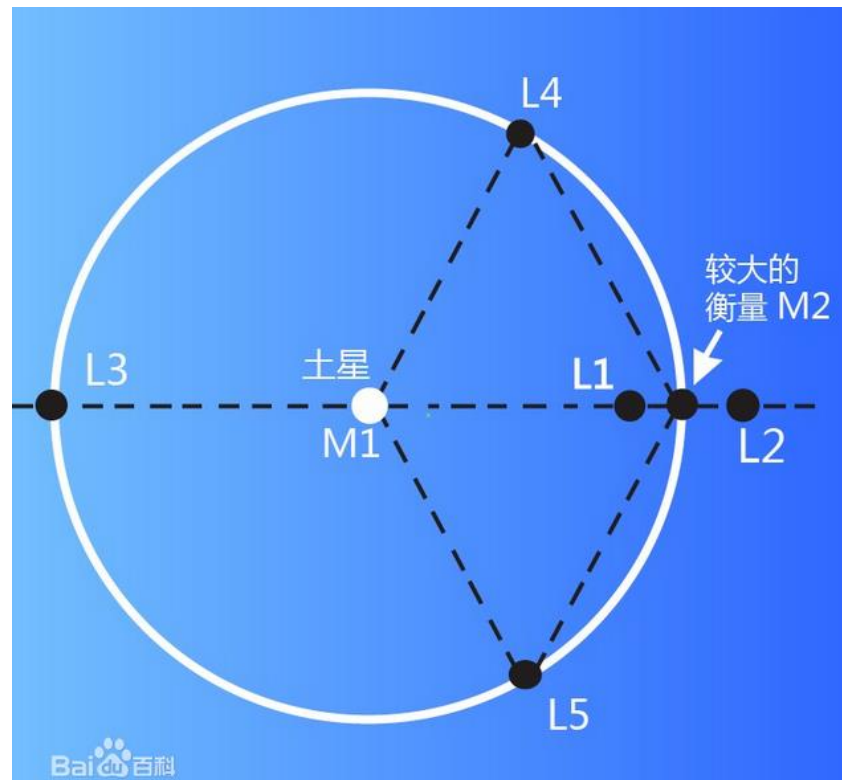
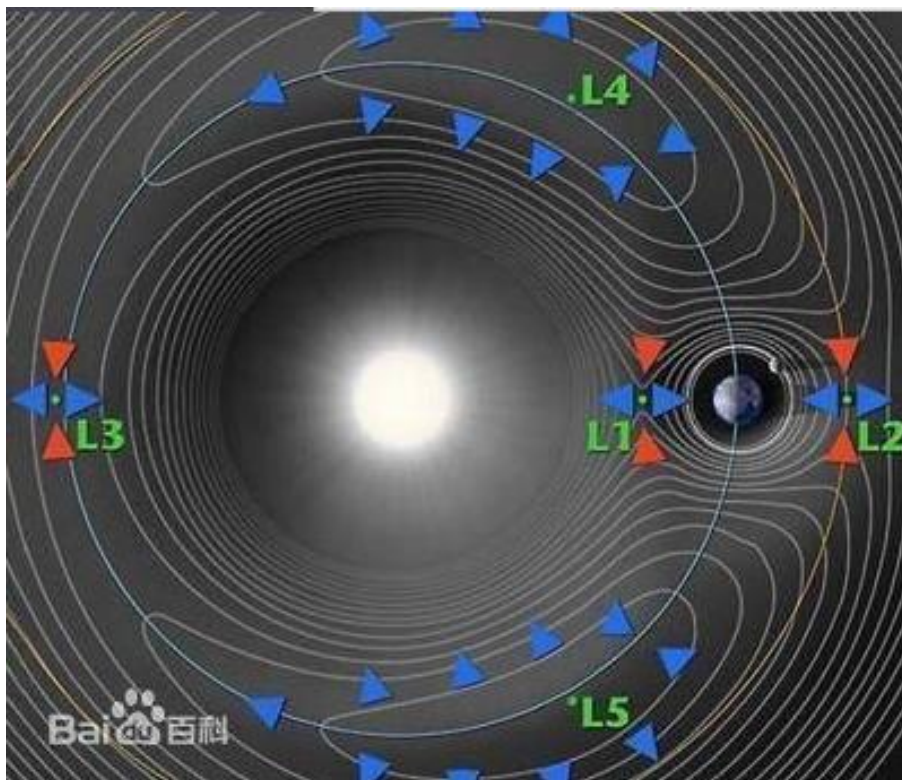
研究生学位课

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- 等式约束的非线性规划

- $\min f(x)$
- $s.t. \quad g_i(x) = 0 \quad i = 1, 2, \dots, m < n$
- 基本思路：将约束问题转化为无约束问题。
- 使用拉格朗日乘子(Lagrangian)
- $L(x, \lambda)$

• 拉格朗日点



- 原问题模型

- $\min f(x)$
 - $s. t. \quad g_i(x) = 0 \quad \forall i = 1, 2, \dots, m < n$

- 拉格朗日函数

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

- 例：6_0

$$\min f(x, y) = x^2 + y^2$$

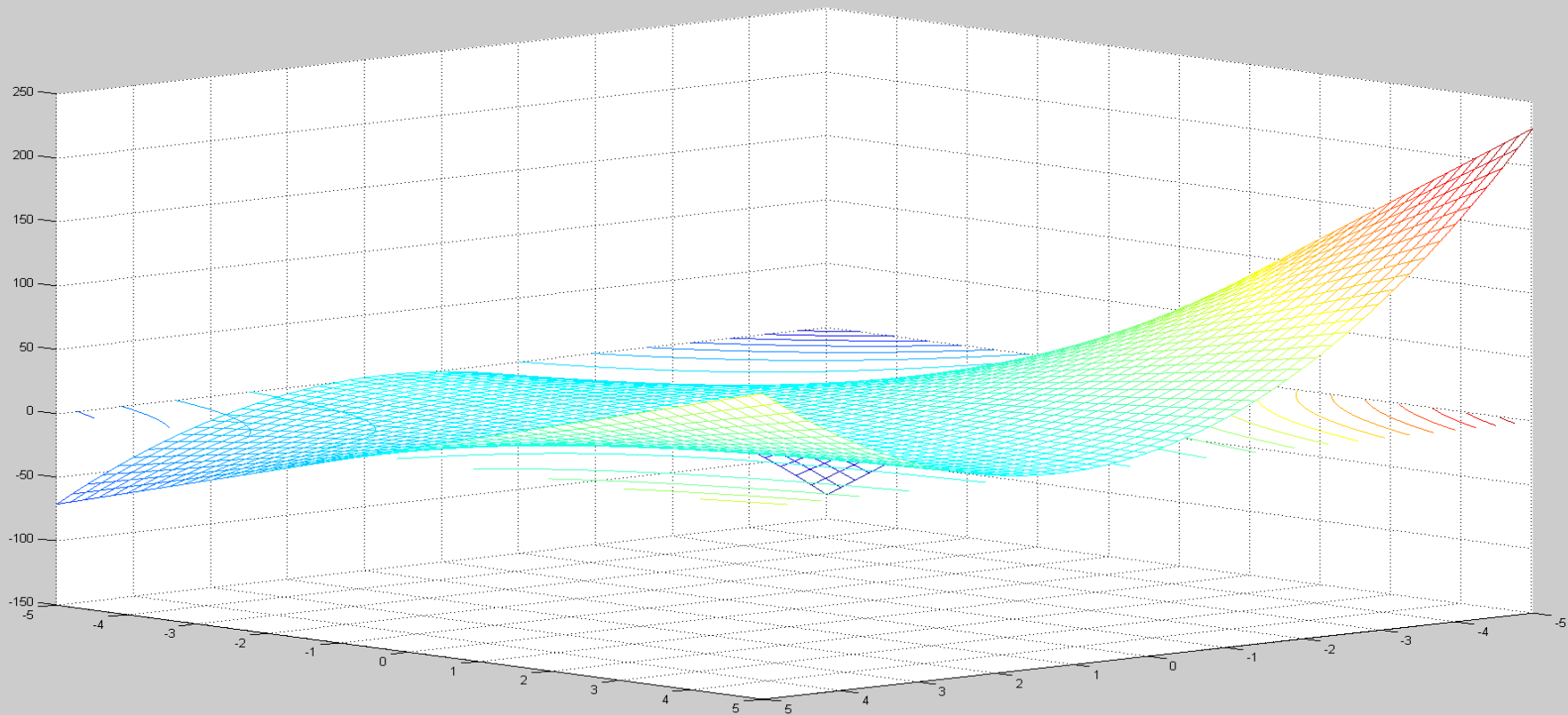
$$s. t. \quad x + y = 1$$

• 例： 6_1

$$\min f(x) = (x - 2)^2$$

$$\text{s.t. } x = 1$$

$$\square L(x, \lambda) = (x - 2)^2 + \lambda(x - 1) \quad x^* = 1, \lambda^* = 2$$



- 等式约束问题解存在的必要性条件

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

- $\frac{\partial L}{\partial x_j} = \frac{\partial L}{\partial x_j} + \sum_{i=1}^m \lambda_i \frac{\partial g_i}{\partial x_j} = 0 \quad j = 1, 2, \dots, n$

- $\frac{\partial L}{\partial \lambda_i} = g_i(x) = 0 \quad i = 1, 2, \dots, m$

- 以上条件成立基础是 无约束问题存在最优解。

- 例6_2

- $\max f(x) = x_1^2 + 4x_2^2$

- $s.t \quad x_1 + 2x_2 = 6$

• 例6_2

$$\square \max f(x) = x_1^2 + 4x_2^2$$

$$\square \text{ s.t. } x_1 + 2x_2 = 6$$

$$\square L(x, \lambda) = x_1^2 + 4x_2^2 + \lambda(x_1 + 2x_2 - 6)$$

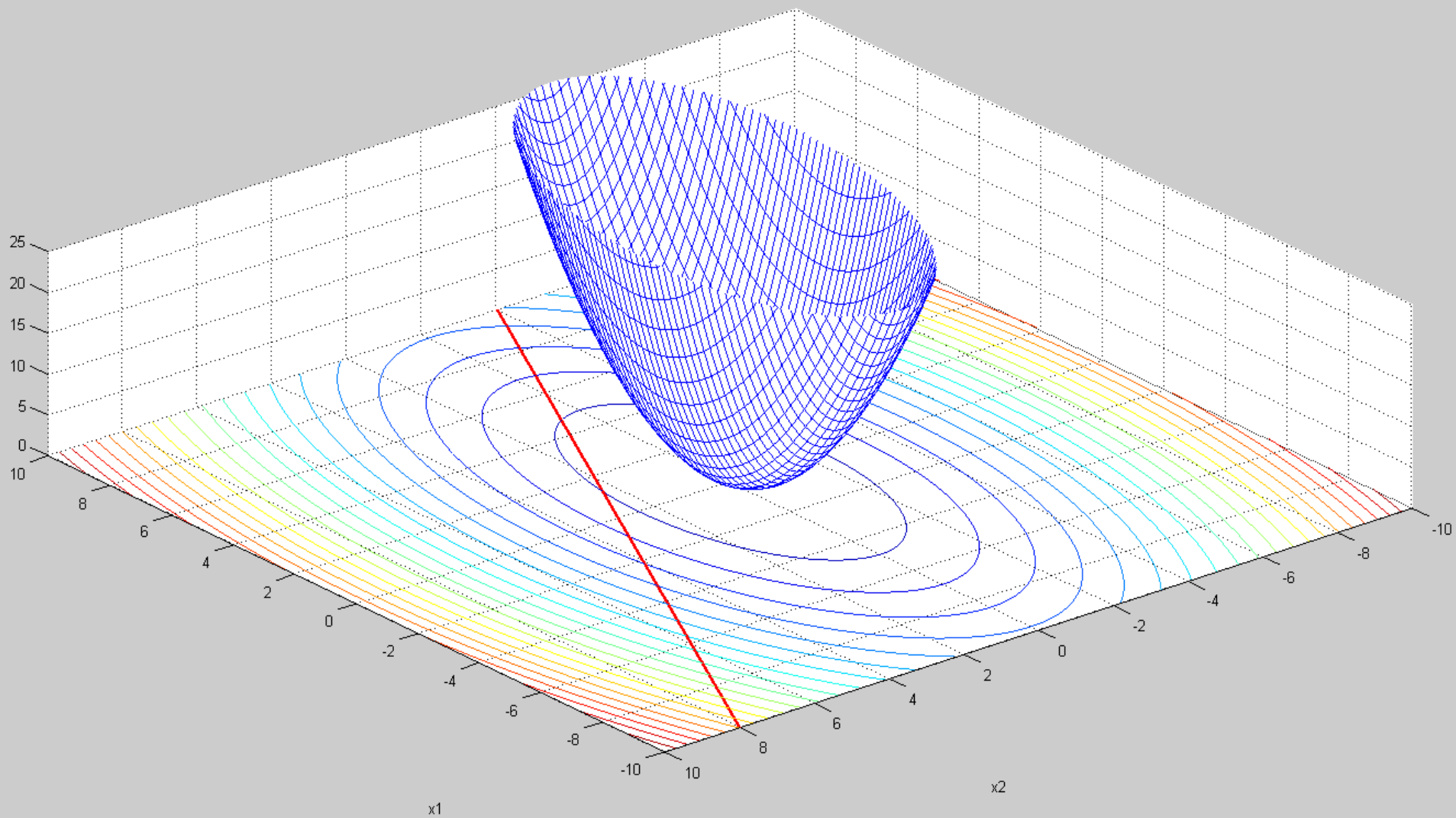
$$\frac{\partial L}{\partial x_1} = 2x_1 + \lambda = 0$$

$$\frac{\partial L}{\partial x_2} = 8x_2 + 2\lambda = 0$$

$$\frac{\partial L}{\partial \lambda} = x_1 + 2x_2 - 6 = 0$$

$$\square \text{ 求得 } x_1^* = 3, x_2^* = \frac{3}{2}, \lambda^* = -6 \quad \text{正确否?}$$

• 例6_2



- 关于约束

- 如果约束条件可以约减，则先约减。
- 加上约束条件后，其最优解一定不优于无约束问题。

$$\begin{aligned} \min f(x) &= x_1^2 + x_2^2 + x_3^2 \\ \text{s.t. } x_1^2 + x_1 x_2 &= 1 \\ x_1 + x_3 &= 4 \end{aligned}$$

- λ 存在性讨论

- 对于Lagrangian 函数

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

及必要性条件:

$$\nabla f(x) + \left[\frac{\partial g}{\partial x} \right] \lambda = 0, \quad g(x) = 0$$

$\left[\frac{\partial g}{\partial x} \right]$ 为 $n \times m$ 矩阵。若找到一个解 x^0 , 满足 $g(x^0) = 0$, 如果此解为最优解, 则 λ 应满足:

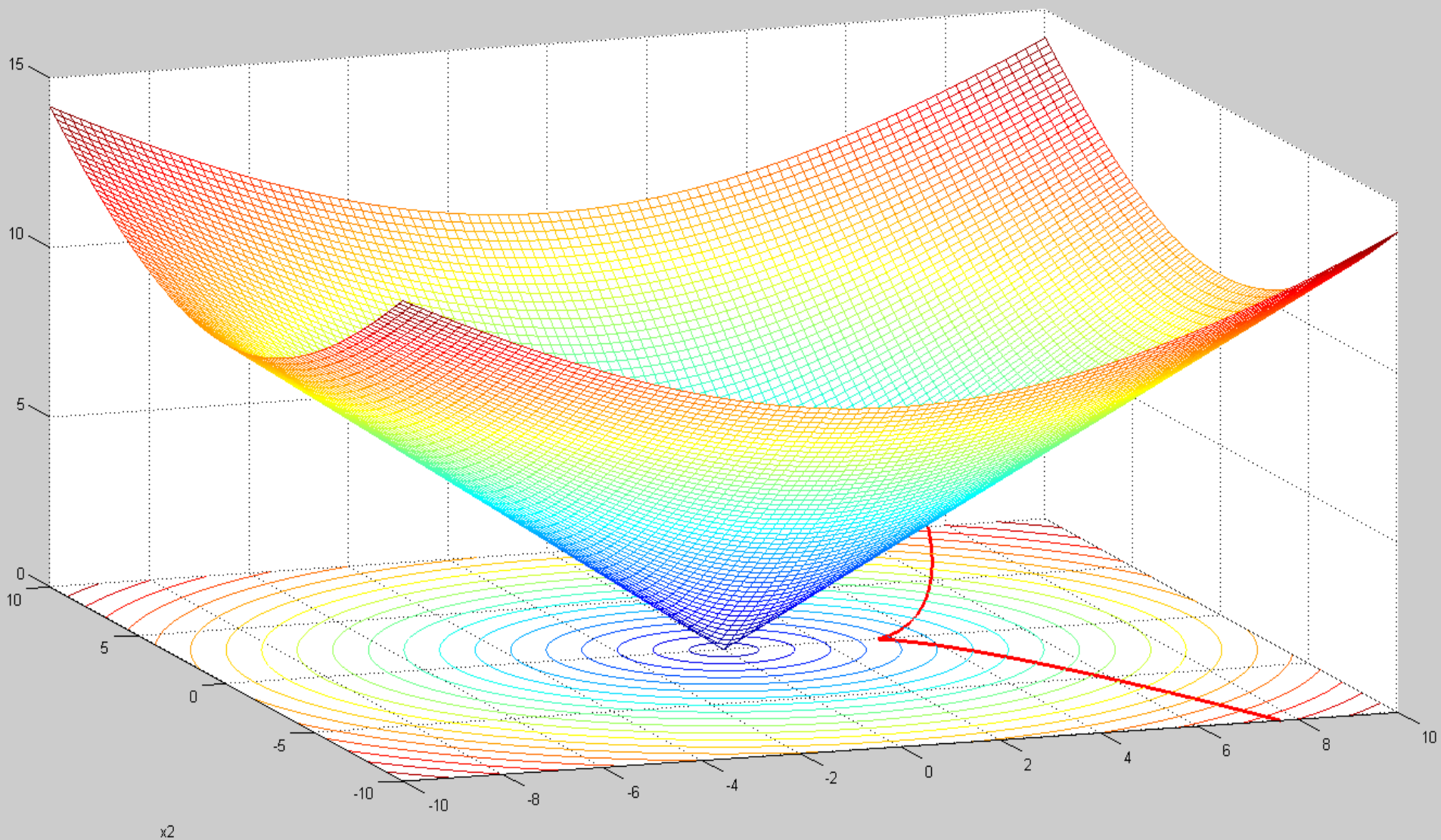
$\left[\frac{\partial g}{\partial x} \right]_{x^0} \lambda = -\nabla f(x^0)$, 此解存在的条件是: $\left[\frac{\partial g}{\partial x} \right]$ 的秩为 m

- 例6_3

$$\min f(x) = \sqrt{x^2 + y^2}$$

$$\text{s.t. } y^2 - (x - 1)^3 = 0$$

• 例6_3



• 例6_3

$$\min f(x) = \sqrt{x^2 + y^2}$$

$$\text{s.t. } y^2 - (x - 1)^3 = 0$$

$$L = \sqrt{x^2 + y^2} + \lambda(y^2 - (x - 1)^3)$$

$$\frac{\partial L}{\partial x} = (x^2 + y^2)^{-1/2} \cdot \frac{1}{2} \cdot 2x + (-3)(x - 1)^2 \lambda = 0$$

$$\frac{\partial L}{\partial y} = (x^2 + y^2)^{-1/2} \cdot \frac{1}{2} \cdot 2y + 2y \lambda = 0$$

$$\frac{\partial L}{\partial \lambda} = y^2 - (x - 1)^3 = 0$$

若将最优点 (1, 0) 代入上式存在问题。Why?

$$\frac{\partial g}{\partial X} = \begin{bmatrix} -3(x - 1)^2 \\ 2y \end{bmatrix} \Big|_{(1,0)} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}, \text{rank}\left(\frac{\partial g}{\partial X}\right)=1$$

- 不等式约束的非线性规划

- $\min f(x)$

- $s. t. \quad h_j(x) \geq 0 \quad j = 1, 2, \dots, m < n$

- 基本思路：将不等式约束问题转化为等式约束问题。

- 使用松弛变量 $\theta_j^2 = h_j(x) \geq 0$

- 拉格朗日函数(不等式约束)

- $L(x, \lambda, \theta) = f(x) + \sum_{j=1}^m \lambda_j (h_j(x) - \theta_j^2)$

- $\frac{\partial L}{\partial x_i} = \frac{\partial f}{\partial x_i} + \sum_{j=1}^m \lambda_j \frac{\partial h_j(x)}{\partial x_i} = 0 \quad i = 1, \dots, n$

- $\frac{\partial L}{\partial \lambda_j} = h_j(x) - \theta_j^2 = 0 \quad j = 1, \dots, m$

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

- 由最后一式可知: $\lambda_j^* = 0$ 或 $\theta_j^* = 0$ 或两者都为0

- 拉格朗日函数(不等式约束)

- 情形I: $\lambda_j^* = 0$ 且 $\theta_j^* \neq 0$, 故 $h_j(x^*) = (\theta_j^*)^2 > 0$, 故最优解在可行域内部。若所有 $\lambda_j^* = 0$, 等价于无约束情形。
- 情形II: $\lambda_j^* \neq 0$ 且 $\theta_j^* = 0$, 故 $h_j(x^*) = 0$, 故最优解在可行域边界。由于 $\lambda_j^* \neq 0$, 故不满足 $\nabla f(x^*) = 0$ 。
- 情形III: $\lambda_j^* = 0$ 且 $\theta_j^* = 0$, 故 $h_j(x^*) = 0$, 且 $\nabla f(x^*) = 0$, 故约束条件的边界穿过无约束问题的最优解。

- 例6_4

- $\min f(x) = (x - a)^2 + b$

- $s.t \quad x \geq c$

• 例6_4

$$\square \min f(x) = (x - a)^2 + b$$

$$s.t \quad x \geq c$$

令 $\theta^2 = x - c \geq 0$, 得到:

$$L = (x - a)^2 + b + \lambda(x - c - \theta^2)$$

$$\square \frac{\partial L}{\partial x} = 2(x - a) + \lambda = 0$$

$$\square \frac{\partial L}{\partial \lambda} = x - c - \theta^2 = 0$$

$$\square \frac{\partial L}{\partial \theta} = -2\lambda\theta = 0$$

• 例6_4

- $2(x - a) + \lambda = 0$
- $x - c - \theta^2 = 0$
- 情形I: $\lambda = 0, x^* = a, \theta^2 = a - c$. 若 $c < a$, θ 为实值.
- 情形II: $\theta = 0, x^* = c, \lambda^* = -2(c - a)$
- 情形III: $\lambda = 0, \theta = 0, x^* = a = c$

- 作业(例6-5)

- 求极大和极小

- $f(x) = 2x^2 - 3y^2 - 2x$

- $s.t \quad x^2 + y^2 \leq 1$

• 作业(6_5)

▣ 求极大和极小

▣ $f(x) = 2x^2 - 3y^2 - 2x$

s.t $x^2 + y^2 \leq 1$

令: $\theta^2 = 1 - x^2 - y^2$

$$L = 2x^2 - 3y^2 - 2x + \lambda(\theta^2 - 1 + x^2 + y^2)$$

▣ $\frac{\partial L}{\partial x} = 4x - 2 + 2\lambda x = 0$

▣ $\frac{\partial L}{\partial y} = -6y + 2\lambda y = 0$

▣ $\frac{\partial L}{\partial \lambda} = \theta^2 - 1 + x^2 + y^2 = 0$

▣ $\frac{\partial L}{\partial \theta} = 2\lambda\theta = 0$

• 作业1

- $1) \theta = 0$

- $4x - 2 + 2\lambda x = 0$

- $-6y + 2\lambda y = 0$

- $-1 + x^2 + y^2 = 0$

解得：

$$\begin{pmatrix} 0.2 \\ \pm\sqrt{0.96} \\ 3 \end{pmatrix}, \quad \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}, \quad \begin{pmatrix} -1 \\ 0 \\ -3 \end{pmatrix}$$

$$-3.2, \quad 0, \quad 4$$

• 作业1

- $2) \lambda = 0$

- $4x - 2 = 0$

- $-6y = 0$

- $-1 + x^2 + y^2 = \theta^2$

解得：

$$\begin{pmatrix} 0.5 \\ 0 \\ \pm\sqrt{1.25} \end{pmatrix}$$

-0.5

• 作业1

- $3) \lambda = 0, \theta = 0$

- $4x - 2 = 0$

- $-6y = 0$

- $-1 + x^2 + y^2 = 0$

无解。

故：极大值为4，极小值为-3.2