

# Ottimizzazione Non Lineare

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# Problemi con soli vincoli di disuguaglianza – 1

Consideriamo il problema

$$\begin{aligned} \min \quad & f(x) \\ \text{c.v.} \quad & g(x) \leq 0 \end{aligned}$$

Usiamo la trasformazione:

- $g_i(x) \leq 0 \quad \rightsquigarrow \quad g_i(x) + y_i^2 = 0, \quad i = 1, \dots, m$

$$L(x, y, \lambda; c) = f(x) + \sum_{i=1}^m \lambda_i (g_i(x) + y_i^2) + \frac{1}{c} \sum_{i=1}^m (g_i(x) + y_i^2)^2$$

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$$\begin{aligned} L_a(x, y, \lambda; \epsilon) &= f(x) + \sum_{i=1}^m \lambda_i (g_i(x) + y_i^2) + \frac{1}{\epsilon} \sum_{i=1}^m (g_i(x) + y_i^2)^2 \\ &= f(x) + \lambda^T g(x) + \frac{1}{\epsilon} \|g(x)\|^2 + \frac{1}{\epsilon} \sum_{i=1}^m (\epsilon \lambda_i y_i^2 + 2g_i(x)y_i^2 + y_i^4) \end{aligned}$$

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abbiamo

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fissati  $\epsilon$  e  $\lambda$ , bisogna “risolvere”

$$\min_{x,y} L_a(x, y, \lambda; \epsilon)$$

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fissati  $\epsilon$  e  $\lambda$ , bisogna “risolvere”

$$\min_{x,y} L_a(x, y, \lambda; \epsilon)$$

# Gradiente della Lagrangiana

$$\nabla_x L_a = \nabla f(x) + \nabla g(x)\lambda + \frac{2}{\epsilon} \sum_{i=1}^m (g_i(x) + y_i^2) \nabla g_i(x)$$

$$\nabla_\lambda L_a = g(x) + y^2$$

$$\nabla_{y_i} L_a = \frac{4}{\epsilon} y_i \left( \epsilon \frac{\lambda_i}{2} + g_i(x) + y_i^2 \right)$$

Quindi, ponendo  $\nabla_{y_i} L_a = 0$  otteniamo

$$y_i \left( \epsilon \frac{\lambda_i}{2} + g_i(x) + y_i^2 \right) = 0$$

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$$y_i^2 = \max \left\{ 0, -\epsilon \frac{\lambda_i}{2} - g_i(x) \right\}$$

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# Espressione per vincoli di disuguaglianza

$$L_a(x, \lambda; \epsilon) = f(x) + \lambda^\top \left( g(x) + \max \left\{ 0, -\epsilon \frac{\lambda}{2} - g(x) \right\} \right) + \frac{1}{\epsilon} \left\| g(x) + \max \left\{ 0, -\epsilon \frac{\lambda}{2} - g(x) \right\} \right\|^2$$

Ovvero

$$L_a(x, \lambda; \epsilon) = f(x) + \lambda^\top \max \left\{ g(x), -\epsilon \frac{\lambda}{2} \right\} + \frac{1}{\epsilon} \left\| \max \left\{ g(x), -\epsilon \frac{\lambda}{2} \right\} \right\|^2$$

$$\nabla_x L_a(x, \lambda; \epsilon) = \nabla f(x) + \nabla g(x) \left( \lambda + \frac{2}{\epsilon} \max \left\{ g(x), -\epsilon \frac{\lambda}{2} \right\} \right)$$

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## Metodo di soluzione - Metodo dei Moltiplicatori

**Algoritmo** SEQLAGR

**Dati:**  $\epsilon_0 > 0$ ,  $\beta > 1$ ,  $\{\tau_k\} \rightarrow 0$ ,  $\rho > 0$ ,  $\lambda_0$ ,  $\mu_0$ , maxit

**for**  $k = 0, 1, \dots, \text{maxit}$

Calcola  $x_k$  t.c.  $\|\nabla_x L_a(x_k, \mu_k, \lambda_k; \epsilon_k)\| \leq \tau_k$

**if**  $\|\nabla L_a(x_k, \mu_k, \lambda_k; \epsilon_k)\| < \rho$  **then**

$x^* \leftarrow x_k$ ,  $\mu^* \leftarrow \mu_k$ ,  $\lambda^* \leftarrow \lambda_k$  STOP

**endif**

$\epsilon_{k+1} = \epsilon_k / \beta$

$$\mu_{k+1} = \mu_k + \frac{2h(x_k)}{\epsilon_k}, \quad \lambda_{k+1} = \lambda_k + \frac{2 \max\{g(x_k), -\epsilon_k \lambda_k / 2\}}{\epsilon_k}$$

**endfor**

**Return:** miglior coppia trovata  $(x^*, \mu^*, \lambda^*)$