

1. 3.1.1

Proposition 3.2.1 (2nd order sufficiency conditions)

Assume that  $f$  and  $h$  are both  $C^2$ .

Let  $x^* \in \mathbb{R}^n$ ,  $\lambda^* \in \mathbb{R}^m$  satisfy  $\nabla_x \mathcal{L}(x^*, \lambda^*) = 0$   $\nabla_\lambda \mathcal{L}(x^*, \lambda^*) = 0$

$$y' \nabla_{xx}^2 \mathcal{L}(x^*, \lambda^*) y > 0 \quad \forall y \neq 0 \text{ with } \nabla h(x^*)' y = 0$$

Then  $x^*$  is a strict local minimum of  $f$  subject to  $h(x) = 0$ .

In fact,  $\exists \delta > 0$  and  $\varepsilon > 0$  s.t.  $f(x) \geq f(x^*) + \frac{\varepsilon}{2} \|x - x^*\|^2 \quad \forall x$  with  $h(x) = 0$  and  $\|x - x^*\| < \delta$

$$(a) \quad f(x) = \|x\|^2 \quad h(x) = \sum_{i=1}^n x_i - 1$$

$$\mathcal{L}(x, \lambda) = f(x) + \lambda h(x) = \sum x_i^2 + \lambda (\sum x_i - 1)$$

By Lagrange multiplier thm,  $\nabla \mathcal{L}(x, \lambda) = 0$  will give us the critical pts.

$$\nabla \mathcal{L}(x, \lambda) = \nabla (\sum x_i^2 + \lambda^* (\sum x_i - 1)) = 0$$

$$2x_i^* + \lambda^* = 0$$

$$x_i^* = -\frac{\lambda^*}{2}$$

$$\text{We have } \sum x_i^* = -\frac{n\lambda^*}{2} = 1 \Leftrightarrow \lambda^* = -\frac{2}{n} \quad x_i = +\frac{2}{n} \cdot \frac{1}{2} = \frac{1}{n}$$

$$\nabla_{xx}^2 \mathcal{L}(x^*, \lambda^*) = \begin{pmatrix} 2 & 0 & \dots & 0 \\ 0 & 2 & & 0 \\ \vdots & & \ddots & \vdots \\ 0 & 0 & & 2 \end{pmatrix} \quad y' \nabla_{xx}^2 \mathcal{L}(x^*, \lambda^*) y = 2 \sum y_i^2 \geq 0 \quad \forall y \in V(x^*)$$

By proposition,  $x^*$  is a local minimum.

$$(b) \quad f(x) = \sum_{i=1}^n x_i \quad h(x) = \|x\|^2 - 1$$

$$\mathcal{L}(x, \lambda) = f(x) + \lambda h(x) = \sum x_i + \lambda (\sum x_i^2 - 1)$$

$$\nabla_x \mathcal{L}(x^*, \lambda^*) = 0$$

$$\nabla (\sum x_i^* + \lambda^* (\sum x_i^{*2} - 1)) = 0$$

$$1 + 2\lambda^* x_i^* = 0$$

$$x_i^* = -\frac{1}{2\lambda^*}$$

$$x_i^{*2} = \frac{1}{4\lambda^{*2}}$$

$$\nabla_{xx}^2 \mathcal{L}(x^*, \lambda^*) = \begin{pmatrix} 2\lambda^* & 0 & \dots & 0 \\ 0 & 2\lambda^* & & 0 \\ \vdots & & \ddots & \vdots \\ 0 & 0 & & 2\lambda^* \end{pmatrix} \quad y' \nabla_{xx}^2 \mathcal{L}(x^*, \lambda^*) y = 2\lambda \sum y_i^2 \geq 0 \quad \forall y \in V(x^*)$$

By proposition,  $x^*$  is a local minimum.

$$(a) \quad \lambda^* = -\frac{\sqrt{n}}{2} \quad x_i^* = \frac{1}{\sqrt{n}} \quad y' \nabla_{xx}^2 \mathcal{L}(x^*, \lambda^*) y = -\sqrt{n} \sum y_i^2 \leq 0$$

2. Industrial Design

A cylindrical can is to hold  $4 \text{ inch}^3$  juice.

The cost per  $\text{inch}^2$  of top + bottom is twice cost / square of side



$$\text{top + bottom area} = 2 \cdot \pi \cdot r^2$$

$$\text{side area} = 2\pi r \cdot h$$

$$\text{minimize } S(r, h) = 2\pi r^2 + 2\pi r h \quad \text{subject to } \pi r^2 \cdot h = 4$$

$$\mathcal{L}(r, h, \lambda) = 2\pi r^2 + 2\pi r h - \lambda(\pi r^2 h - 4)$$

$$\nabla \mathcal{L}(r, h, \lambda) = \vec{0} \quad \text{solve,}$$

$$r = \sqrt[3]{\frac{2}{\pi}} \quad h = 2\sqrt[3]{\frac{2}{\pi}} \quad \lambda = 2^{2/3} \pi^{1/3}$$

### 3. Duality

$$\min_{A'x \geq b} c'x \Leftrightarrow \max_{A\mu = c, \mu \geq 0} b'\mu$$

Proof: ( $\Rightarrow$ )

the dual function is given by

$$\begin{aligned} g(\mu) &= \min_{x \in \mathbb{R}^n} \{f(x) + \mu g(x)\} = \min_{x \in \mathbb{R}^n} \{c'x + \mu(b - A'x)\} = \min_{x \in \mathbb{R}^n} \{(c' - \mu A')x + \mu b\} \\ &= \begin{cases} b'\mu & \text{if } c = \mu A \\ -\infty & \end{cases} \end{aligned}$$

$$\Leftrightarrow \max_{\mu \geq 0} g(\mu) \text{ subject to } \mu \geq 0 \text{ or equivalently } \max_{A\mu = c, \mu \geq 0} b'\mu$$

( $\Leftarrow$ )

$$\begin{aligned} p(x) &= \max_{\mu \geq 0} \{f(\mu) + x g(\mu)\} = \max_{\mu \geq 0} \{b'\mu + x(c' - A'\mu)\} = \max_{\mu \geq 0} \{(b' - A'x)\mu + c'x\} \\ &= \begin{cases} c'x & \text{if } b - A'x \leq 0 \\ \infty & \end{cases} \end{aligned}$$

4. 4.2.1

$$\text{minimize } f(x) = \frac{1}{2}(x_1^2 - x_2^2) - 3x_2 \quad \text{subject to } x_2 = 0$$

(a) Calculate optimal solution and LM.

$$\begin{aligned} \mathcal{L}(x, \lambda) &= \frac{1}{2}(x_1^2 - x_2^2) - 3x_2 + \lambda x_2 \\ &= \frac{1}{2}x_1^2 - \frac{1}{2}x_2^2 + (\lambda - 3)x_2 \end{aligned}$$

$$\frac{\partial \mathcal{L}}{\partial x_1} = x_1 = 0$$

$$\frac{\partial \mathcal{L}}{\partial x_2} = -x_2 + \lambda - 3 = 0$$

$$\bar{x}^* = (0, 0) \quad \lambda^* = 3$$

(b) For  $k=0, 1, 2, \dots$  and  $c^k = 10^{k+1}$ , calculate and compare the iterates of the quadratic penalty method with  $\lambda^k = 0 \quad \forall k$  and multiplier with  $\lambda^0 = 0$

$$\mathcal{L}(x, \lambda) = \frac{1}{2}(x_1^2 - x_2^2) - 3x_2 + \lambda x_2 + \frac{c}{2}x_2^2$$

$$\frac{\partial \mathcal{L}}{\partial x_1^k} = x_1^k = 0$$

$$\frac{\partial \mathcal{L}}{\partial x_2^k} = -x_2^k - 3 + \lambda^k + c^k x_2^k = 0 \quad x_2^k = \frac{3 - \lambda^k}{c^k - 1}$$

$$\textcircled{1} \quad x^k = \left(0, \frac{3 - \lambda^k}{c^k - 1}\right)$$

$$c^k = 10^{k+1}$$

$$k=0 \Rightarrow c^0 = 10^{0+1} = 10$$

$$x^0 = \left(0, \frac{1}{3}\right)$$

$$k=1 \quad c^1 = 100$$

$$x^1 = \left(0, \frac{1}{33}\right)$$

$$k=2 \quad c^2 = 1000$$

$$x^2 = \left(0, \frac{1}{333}\right)$$

⑤ multiplier method with  $\lambda^0 = 0$  let  $h(x) = \frac{3 - \lambda^k}{c^k - 1}$

$$k=0 \Rightarrow c^0 = 10, \lambda^0 = 0 \quad x^0 = \left(0, \frac{1}{3}\right)$$

$$k=1 \Rightarrow c^1 = 100 \quad \lambda^1 = \lambda^0 + c^0 h(x_1^0) = 0 + 10 \cdot \frac{1}{3} = \frac{10}{3}$$

$$x_2^1 = \frac{3 - \frac{10}{3}}{100 - 1} = -\frac{1}{297}$$

$$k=2 \Rightarrow c^2 = 1000$$

$$\lambda^2 = \lambda^1 + c^1 h(x_2^1) = \frac{10}{3} + 100 \cdot \left(-\frac{1}{297}\right) = 3$$

$$x_2^2 = 0 \Rightarrow \bar{x}_2 = (0, 0)$$

(c) Suppose  $c$  is taken to be constant in the method of multipliers. For what values of  $c$  would the augmented Lagrangian have a minimum and for what values of  $c$  would the method converge.

$$\begin{aligned} \lambda^{k+1} &= \lambda^k + c^k h(x_2^k) = \lambda^k + c^k \left(\frac{3 - \lambda^k}{c^k - 1}\right) = \frac{3c^k - \lambda^k}{c^k - 1} \\ \lambda^{k+1} - \lambda^k &= \frac{3c^k - \lambda^k}{c^k - 1} - \lambda^k = \frac{3c^k - \lambda^k - 3c^k + 3}{c^k - 1} = \frac{3 - \lambda^k}{c^k - 1} \end{aligned}$$

So min exist when  $c^k > 1$ .

method converge when  $\frac{1}{c^k - 1} < 1 \Rightarrow c^k > 2$ .