



$$\min. f(x) \quad \text{the optimal is } x^*$$

$$\text{s.t. } h_i(x) = 0$$

$$L(x, v) = f(x) + \sum_i v_i h_i(x)$$

$$A(x, v, c) = L(x, v) + \frac{c}{2} \sum_i \|h_i(x)\|_2^2 = f(x) + \sum_i v_i h_i(x) + \frac{c}{2} \sum_i \|h_i(x)\|_2^2$$

$$\text{optimality condition: } \left. \begin{aligned} h_i(x^*) &= 0 \\ \nabla L(x^*, v^*) &= \nabla f(x^*) + \sum_i v_i^* \nabla h_i(x^*) \end{aligned} \right\} \begin{aligned} &\text{both } L(x, v) \\ &\text{and } A(x, v, c) \text{ is minimized} \\ &\text{at } (x^*, v^*) \end{aligned}$$

$$\nabla_x A(x^*, v^*, c) = \nabla L(x^*, v^*) + c \sum_i h_i(x^*) \cdot \nabla h_i(x^*) = 0$$

$$\text{we want } \min_{x, v} A(x, v, c) \text{ while keep } \nabla_x A(x, v, c) = \nabla_x L(x, v)$$

$$x' = \arg \min_x A(x, v, c)$$

$$\left. \begin{aligned} \nabla_x A(x', v, c) &= \nabla f(x') + \sum_i v_i \nabla h_i(x') + c \sum_i h_i(x') \cdot \nabla h_i(x') \\ \nabla_x L(x', v', c) &= \nabla f(x') + \sum_i v'_i \nabla h_i(x') \end{aligned} \right\}$$

$$v'_i = v_i + c \cdot h_i(x')$$

eg. least-norm solution

$$\begin{aligned} \min. \quad & x^T x \\ \text{s.t.} \quad & Ax = b \end{aligned} \quad \begin{aligned} L(x, v) &= x^T x + v^T (Ax - b) \\ A(x, v, c) &= x^T x + v^T (Ax - b) + \frac{c}{2} \|Ax - b\|_2^2 \end{aligned}$$

$$x' = \arg \min_x A(x, v, c)$$

$$\nabla_x A(x', v, c) = 2x' + A^T v + c \cdot A^T (Ax - b) = 0$$

$$2x' + A^T v + c \cdot A^T Ax - c \cdot A^T b = 0$$

$$(2I + c \cdot A^T A) x' = A^T (c \cdot b - v)$$

$$\nabla_x A(x', v, c) = \nabla L(x', v')$$

$$\nabla_x A(x', v, c) = 2x' + A^T v + c \cdot A^T (Ax' - b)$$

$$\nabla_x L(x', v') = 2x' + A^T v'$$

$$v' = v + c \cdot (Ax' - b)$$