

# Quadratic Programming

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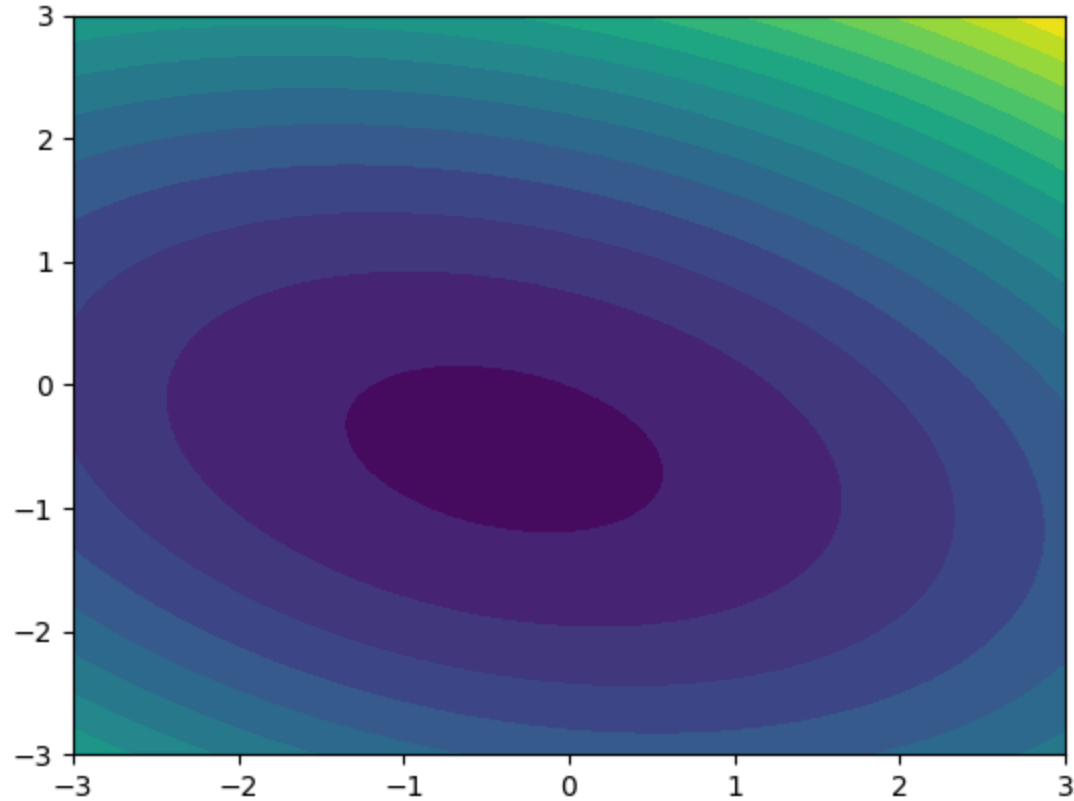
# Quadratic programming

- Quadratic programming (QP) is the process of solving certain mathematical optimization problems involving quadratic functions subject to linear constraints on the variables.

$$\begin{aligned} \min_x \quad & \frac{1}{2}x^T Px + q^T x \\ \text{subject to} \quad & Ax = b \\ & Gx \leq h \end{aligned}$$

# Unconstrained QP

- $\min_x \frac{1}{2}x^T P x + q^T x$



# Unconstrained QP

- if  $x_* = \arg \min_x \frac{1}{2}x^T P x + q^T x$ , then  $\frac{\partial}{\partial x} (\frac{1}{2}x_*^T P x_* + q^T x_*) = 0$
- Using  $\frac{\partial}{\partial x} (\frac{1}{2}x^T P x + q^T x) = P x + q$ :

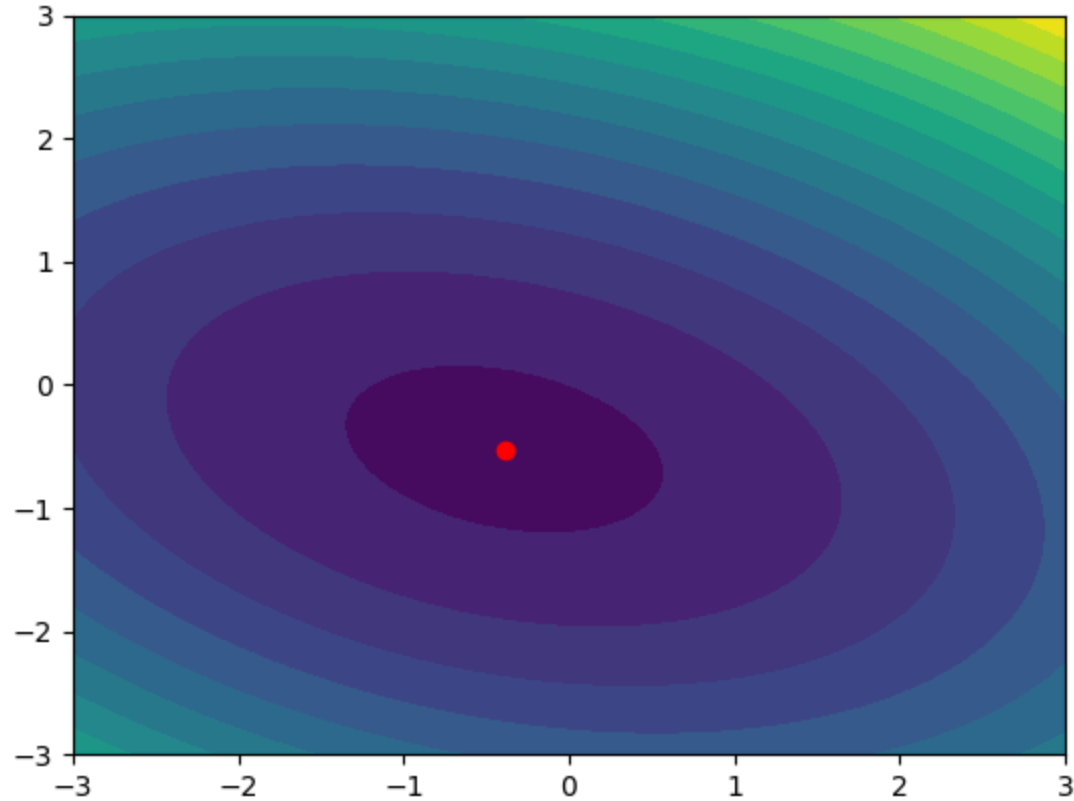
$$P x_* + q = 0$$

$$P x_* = -q$$

$$x_* = -P^{-1}q$$

# Unconstrained QP

- $x_* = -P^{-1}q$



# Lagrangian multiplier

- Consider the optimization problem:

$$\min_x f(x) \quad \text{s. t. } Ax = b; \quad Gx \leq h$$

- Lagrangian multiplier  $\mathcal{L}$ :

$$\mathcal{L}(x, u, v) = f(x) + u^T (Ax - b) + v^T (Gx - h)$$

- The optimization problem above can be rewritten as follows:

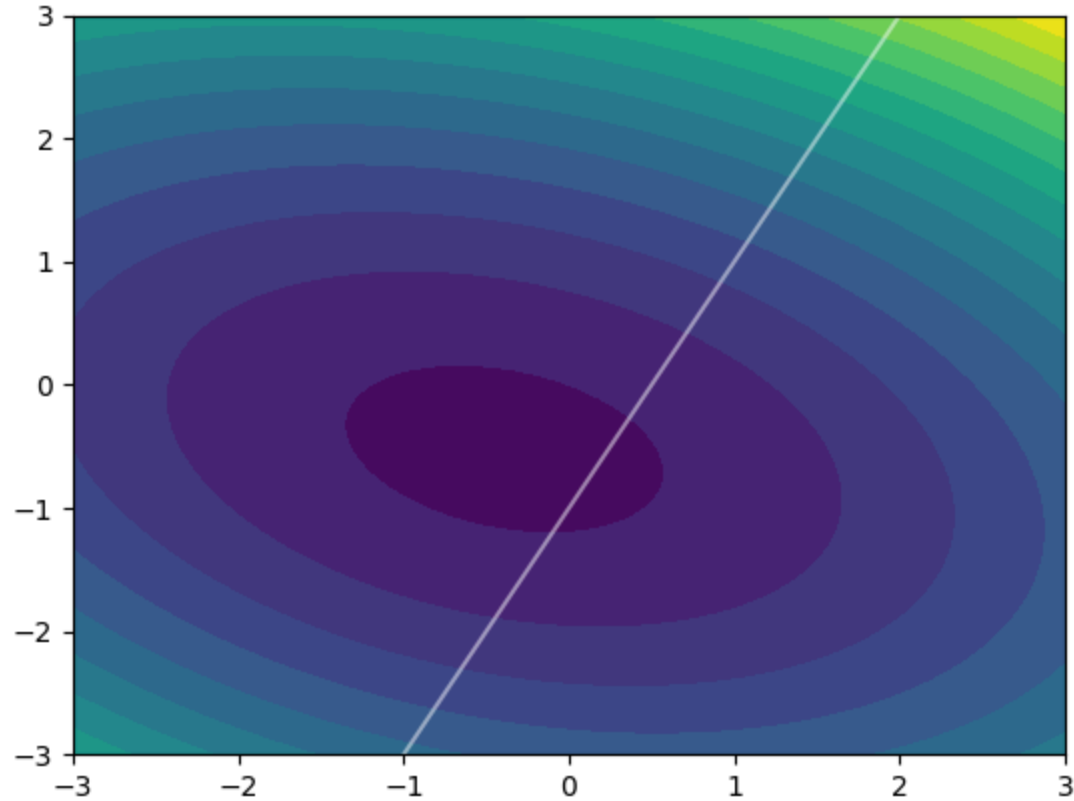
$$\max_{u,v} \min_x \mathcal{L}(x, u, v) \quad \text{s. t. } v \geq 0$$

- We call  $g(u, v) = \min_x \mathcal{L}(x, u, v)$  as a dual function, and Lagrangian dual problem is:

$$\max_{u,v} g(u, v) \quad \text{s. t. } v \geq 0$$

# Equality constrained QP

- $\min_x \frac{1}{2}x^T P x + q^T x \quad \text{s. t. } Ax = b$



# Equality constrained QP

- Dual function:  $g(u) = \min_x \mathcal{L}(x, u) = \frac{1}{2}x^T P x + q^T x + u^T (Ax - b)$
- Using  $\frac{\partial}{\partial x} \mathcal{L}(x, u) = Px + q + A^T u$ ,  $\arg \min_x \mathcal{L}(x, u) = -P^{-1}(A^T u + q)$ . then:

$$\begin{aligned} g(u) &= \frac{1}{2}(A^T u + q)^T P^{-1}(A^T u + q) - q^T P^{-1}(A^T u + q) - u^T (AP^{-1}(A^T u + q) + b) \\ &= -\frac{1}{2}(u^T A^T P^{-1} A^T u + q^T P^{-1} q) - (AP^{-1}q + b)^T u \end{aligned}$$

- Using  $\frac{\partial g(u)}{\partial u} = -(A^T P^{-1} A)u - (AP^{-1}q + b)$ :

$$u_* = -(A^T P^{-1} A)^{-1}(AP^{-1}q + b)$$

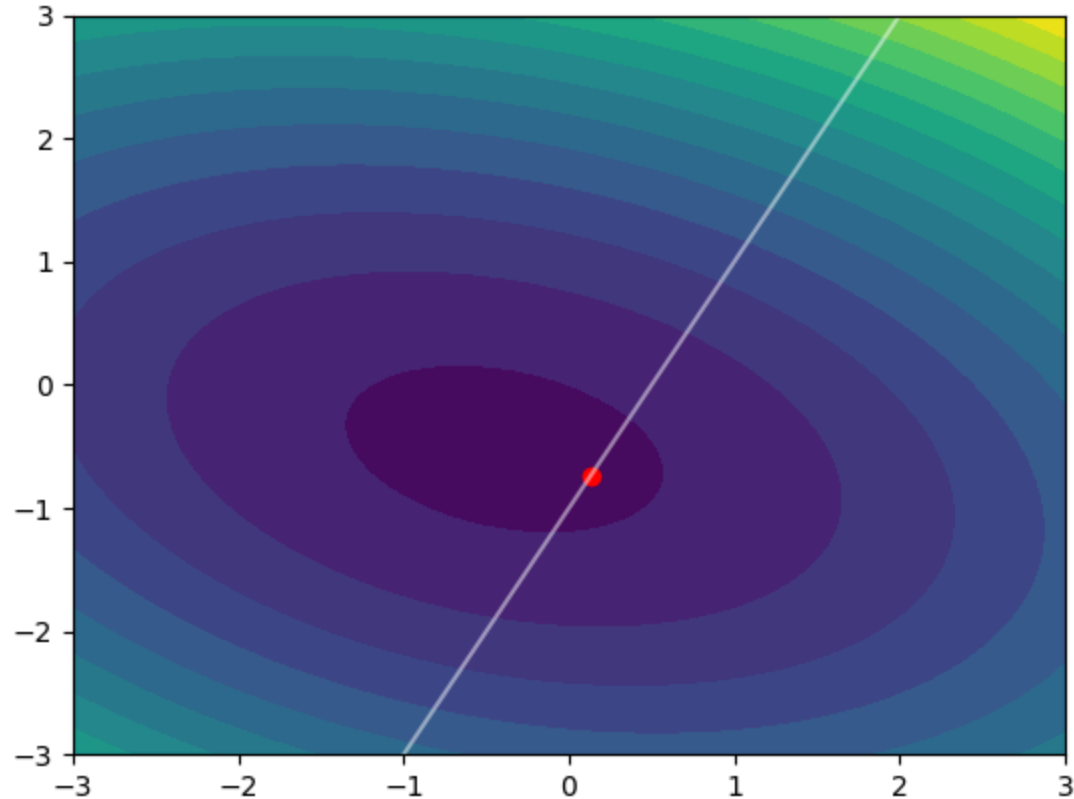
$$x_* = -P^{-1}(A^T u_* + q)$$

$$= P^{-1}(A^T (A^T P^{-1} A)^{-1}(AP^{-1}q + b) - q)$$



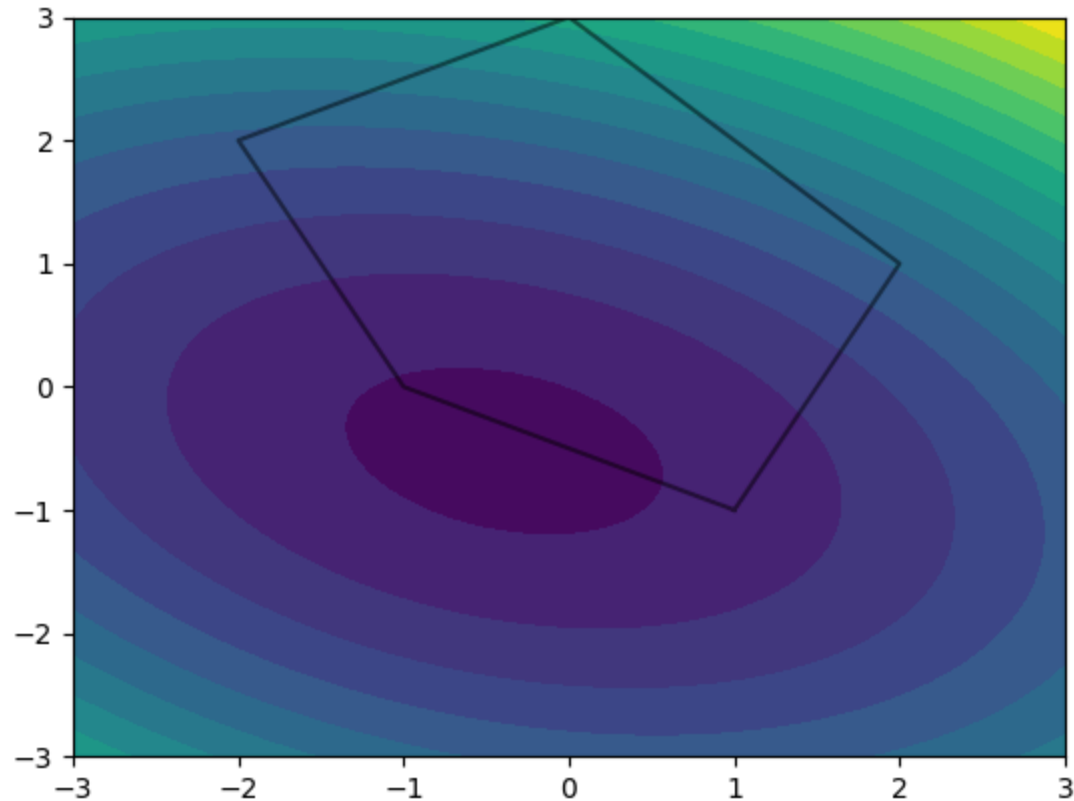
# Equality constrained QP

- $x_* = P^{-1}(A^T(A^T P^{-1} A)^{-1}(AP^{-1}q + b) - q)$



# Inequality constrained QP

- $\min_x \frac{1}{2}x^T Px + q^T x \quad \text{s. t. } Gx \leq h,$



# Inequality constrained QP

- Penalty method

$$\min_x \frac{1}{2}x^T Px + q^T x + \sum_{i=0}^n \max(0, g_i^T x - h)^2$$

- Barrier method

$$\min_x \frac{1}{2}x^T Px + q^T x + \frac{1}{t}\phi(x), \text{ where } \phi(x) := \begin{cases} \sum_{i=0}^n -\log(h - g_i^T x) & \text{for } Gx < h \\ \infty & \text{otherwise} \end{cases}$$

- Primal-dual interior point method
- Active-set method

$$\min_x \frac{1}{2}x^T Px + q^T x \quad \text{s. t. } g_i^T x = h_i, \forall i \in \{i | g_i^T x > h_i, i = 1, \dots, n\}$$

# Inequality constrained QP (Active-set method)

- Find unconstrained problem solution  $x_{\text{unc}} = \arg \min_x \frac{1}{2}x^T P x + q^T x$

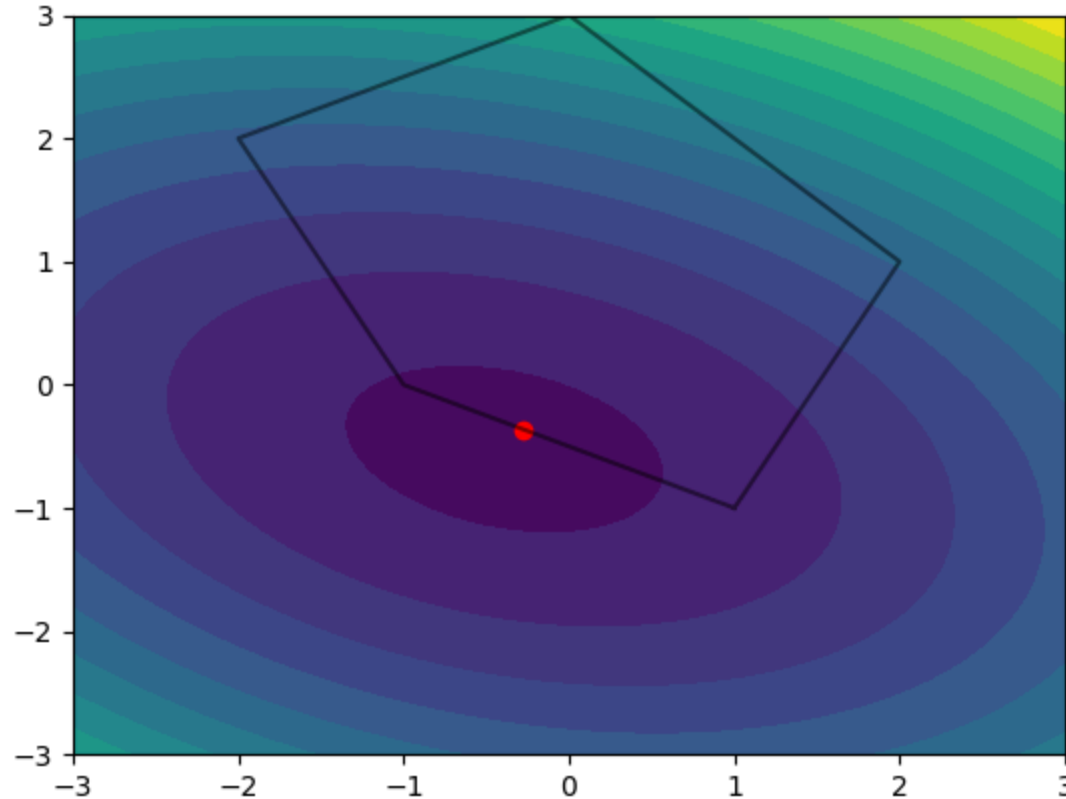
- Using  $G = \begin{bmatrix} \vdots \\ g_i^T \\ \vdots \end{bmatrix}$ ,  $h = \begin{bmatrix} \vdots \\ h_i \\ \vdots \end{bmatrix}$  and  $x_{\text{unc}}$ , find active inequality constraints:

$$G_a = \begin{bmatrix} \vdots \\ g_i^T \\ \vdots \end{bmatrix}, h_a = \begin{bmatrix} \vdots \\ h_i \\ \vdots \end{bmatrix}, \forall i \in \{i | g_i^T x_{\text{unc}} > h_i, i = 1, \dots, n\}$$

- Solve equality constrained problem  $\min_x \frac{1}{2}x^T P x + q^T x \quad \text{s. t. } G_a x = h_a$

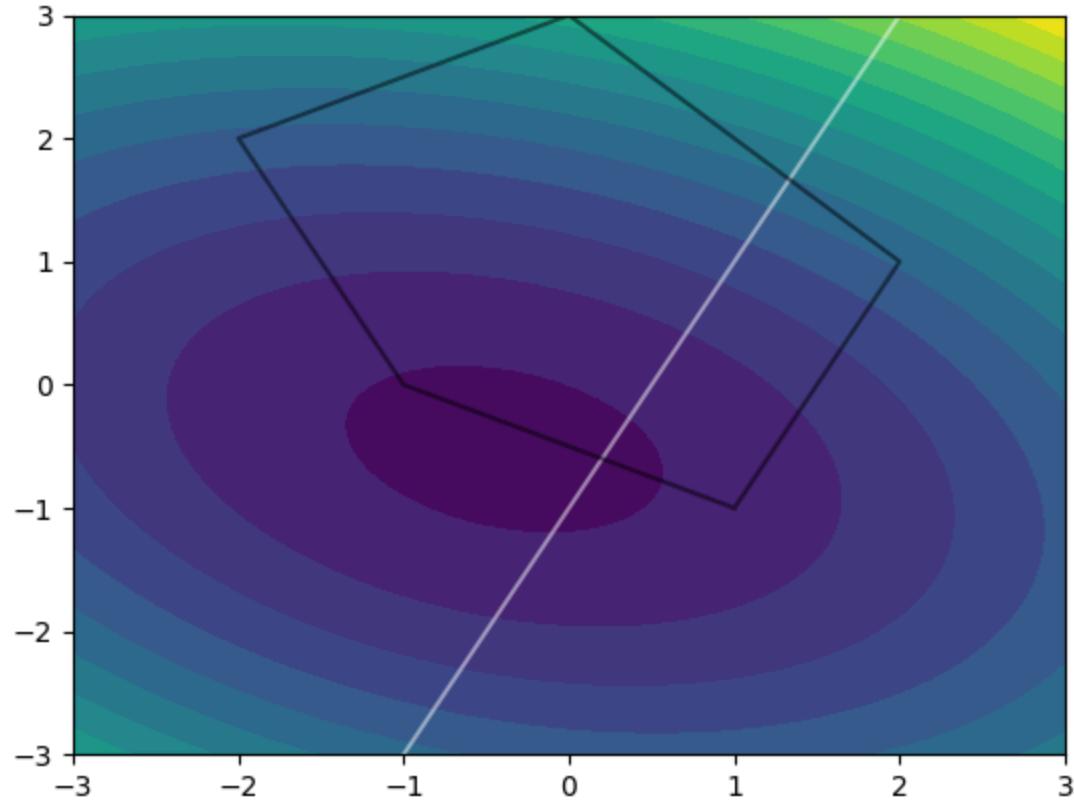
# Inequality constrained QP (Active-set method)

- $x_* = \arg \min_x \frac{1}{2}x^T P x + q^T x \quad \text{s. t. } G_a x = h_a$



# General QP

- $\min_x \frac{1}{2}x^T Px + q^T x \quad \text{s. t. } Ax = b; Gx \leq h$



# General QP

- Penalty method

$$\min_x \frac{1}{2}x^T Px + q^T x + \sum_{i=0}^n \max(0, g_i^T x - h)^2 \quad \text{s. t. } Ax = b$$

- Barrier method

$$\min_x \frac{1}{2}x^T Px + q^T x + \frac{1}{t}\phi(x) \quad \text{s. t. } Ax = b$$

- Primal-dual interior point method

- Active-set method

$$\min_x \frac{1}{2}x^T Px + q^T x \quad \text{s. t. } \begin{bmatrix} A \\ G_a \end{bmatrix} x = \begin{bmatrix} b \\ h_a \end{bmatrix}$$

# General QP

- $x_* = \begin{bmatrix} A \\ G_a \end{bmatrix}^{-1} \begin{bmatrix} b \\ h_a \end{bmatrix}$  (active-set method)

