

Quadratic Optimization: Primal Active-Set Algorithm

Consider the quadratic optimization problem

$$\begin{aligned} & \text{minimize} && \frac{1}{2} \mathbf{x}^T Q \mathbf{x} + \mathbf{c}^T \mathbf{x} + \beta \\ & \text{subject to} && A \mathbf{x} \geq \mathbf{b} \end{aligned} \tag{1}$$

where $Q \succ 0$.

Below, we reproduce the primal active-set method (the motivation for each step was discussed in class) for optimization.

Step 1

Input a feasible point, \mathbf{x}^0 , identify the active set \mathcal{I}^0 , form matrix $A_{\mathcal{I}^0}$, and set $k = 0$.

Step 2

Compute $\mathbf{g}^k = Q\mathbf{x}^k + \mathbf{c}$.

Check the rank condition $\text{rank}[A_{\mathcal{I}^k}^T \quad \mathbf{g}^k] = \text{rank}[A_{\mathcal{I}^k}^T]$. If it does not hold, go to **Step 4**.

Step 3

Solve the system $A_{\mathcal{I}^k}^T \hat{\lambda} = \mathbf{g}^k$. If $\hat{\lambda} \geq \mathbf{0}$, output \mathbf{x}^k as the solution and stop; otherwise, remove the index that is associated with the most negative Lagrange multiplier (some $\hat{\lambda}_t$) from \mathcal{I}^k .

Step 4

Compute the value of \mathbf{d}^k :

$$\begin{aligned} \mathbf{d}^k = & \underset{\mathbf{d}}{\text{argmin}} && \frac{1}{2} \mathbf{d}^T Q \mathbf{d} + (\mathbf{g}^k)^T \mathbf{d} \\ & \text{subject to} && \mathbf{a}_i^T \mathbf{d} = 0 \quad \text{for } i \in \mathcal{I}^k \end{aligned} \quad (2)$$

Step 5

Compute α_k :

$$\alpha_k = \min \left\{ 1, \min_{\substack{j \notin \mathcal{I}^k \\ \mathbf{a}_j^T \mathbf{d}^k < 0}} \frac{\mathbf{a}_j^T \mathbf{x}^k - b_j}{-\mathbf{a}_j^T \mathbf{d}^k} \right\} \quad (3)$$

Set $\mathbf{x}^{k+1} = \mathbf{x}^k + \alpha_k \mathbf{d}^k$.

Step 6

If $\alpha_k < 1$, construct \mathcal{I}^{k+1} by adding the index that yields the minimum value of α_k in (3). Otherwise, let $\mathcal{I}^{k+1} = \mathcal{I}^k$.

Step 7

Set $k = k + 1$ and repeat from **Step 2**.

Figure 1: Optimization for the quadratic problem in (??) using Primal Active-set Method.