

Another Approach to the Problem in (4.7)

$$\begin{aligned}
& \underset{\mathbf{x}, y, \mathbf{z}}{\text{minimize}} && \sum_{i=1}^m (z_i - y - d_i)^2 \\
& \text{subject to:} && \|\mathbf{x} - \mathbf{a}_i\| = z_i, \quad i = 1, 2, \dots, m \\
& && \|\mathbf{x}\| = y
\end{aligned} \tag{4.7a-c}$$

In the k th iteration, the k th iterate $\{\mathbf{x}_k, y_k, \mathbf{z}_k\}$. Let

$$\begin{aligned}
\mathbf{x}_{k+1} &= \mathbf{x}_k + \boldsymbol{\delta}_x \\
y_{k+1} &= y_k + \delta_y \\
\mathbf{z}_{k+1} &= \mathbf{z}_k + \boldsymbol{\delta}_z
\end{aligned} \tag{4.8a-c}$$

where $\{\boldsymbol{\delta}_x, \delta_y, \boldsymbol{\delta}_z\}$ are such that the constraints in (4.7b) and (4.7c) are better approximated at $\{\mathbf{x}_{k+1}, y_{k+1}, \mathbf{z}_{k+1}\}$ in the sense that

$$\begin{aligned}
\|\mathbf{x}_{k+1} - \mathbf{a}_i\| &\approx z_i^{(k+1)}, \quad i = 1, 2, \dots, m \\
\|\mathbf{x}_{k+1}\| &\approx y_{k+1}
\end{aligned}$$

namely,

$$\begin{aligned}
\|\mathbf{x}_k + \boldsymbol{\delta}_x - \mathbf{a}_i\| &\approx z_i^{(k)} + \delta_{z_i}, \quad i = 1, 2, \dots, m \\
\|\mathbf{x}_k + \boldsymbol{\delta}_x\| &\approx y_k + \delta_y
\end{aligned}$$

By replacing the left-hand sides of the above equations with their first-order Taylor approximations, we obtain

$$\begin{aligned}
\|\mathbf{x}_k - \mathbf{a}_i\| + \partial^T \|\mathbf{x}_k - \mathbf{a}_i\| \boldsymbol{\delta}_x &\approx z_i^{(k)} + \delta_{z_i}, \quad i = 1, 2, \dots, m \\
\|\mathbf{x}_k\| + \partial_x^T \|\mathbf{x}_k\| \boldsymbol{\delta}_x &\approx y_k + \delta_y
\end{aligned}$$

where ∂_x is the subdifferential operator with respect to variable \mathbf{x} . Assuming $\mathbf{x}_k \neq \mathbf{a}_i$ and \mathbf{x}_k is nonzero, then

$$\partial_x \|\mathbf{x}_k - \mathbf{a}_i\| = \frac{\mathbf{e}}{\|\mathbf{x}_k - \mathbf{a}_i\|} \quad \text{and} \quad \partial_x \|\mathbf{x}_k\| = \frac{\mathbf{e}}{\|\mathbf{x}_k\|}$$

where \mathbf{e} is the all-one vector. Hence

$$\| \mathbf{x}_k - \mathbf{a}_i \| + \frac{\mathbf{e}^T \boldsymbol{\delta}_x}{\| \mathbf{x}_k \|} \approx z_i^{(k)} + \delta_{z_i}, \quad i = 1, 2, \dots, m \quad (4.9a)_{\text{new}}$$

$$\| \mathbf{x}_k \| + \frac{\mathbf{e}^T \boldsymbol{\delta}_x}{\| \mathbf{x}_k \|} \approx y_k + \delta_y \quad (4.9b)_{\text{new}}$$

Based on this, the problem to be solved in the k th iteration is formulated as

$$\begin{aligned} & \underset{\tilde{\boldsymbol{\delta}}}{\text{minimize}} \quad f(\tilde{\boldsymbol{\delta}}) = \sum_{i=1}^m (-\delta_y + \delta_{z_i} - \tilde{d}_i^{(k)})^2 \\ & \text{subject to:} \quad \| \mathbf{x}_k - \mathbf{a}_i \| + \frac{\mathbf{e}^T \boldsymbol{\delta}_x}{\| \mathbf{x}_k \|} = z_i^{(k)} + \delta_{z_i}, \quad i = 1, 2, \dots, m \\ & \quad \quad \quad \| \mathbf{x}_k \| + \frac{\mathbf{e}^T \boldsymbol{\delta}_x}{\| \mathbf{x}_k \|} = y_k + \delta_y \\ & \quad \quad \quad \begin{bmatrix} -\beta \mathbf{1}_2 \\ -\min\{\beta, y_k\} \\ -\min\{\beta, z_k\} \end{bmatrix} \leq \begin{bmatrix} \boldsymbol{\delta}_x \\ \delta_y \\ \boldsymbol{\delta}_z \end{bmatrix} \leq \begin{bmatrix} \beta \mathbf{1}_2 \\ \beta \\ \beta \mathbf{1}_m \end{bmatrix} \end{aligned} \quad (4.10a-d)_{\text{new}}$$

The constraints in (4.10d) not only assure that the magnitude of each component in $\{\boldsymbol{\delta}_x, \delta_y, \boldsymbol{\delta}_z\}$ is no greater than β , but also they assure that all components of $\{y_{k+1}, z_{k+1}\}$ are nonnegative as long as $\{y_k, z_k\}$ are nonnegative, which are natural to impose as can be seen from (4.7b) and (4.7c) because they are vector norms. Obviously, the problem in (4.10) is a convex QP problem. One technical difficulty that may occur in solving problem (4.10) is that the feasible region defined by (4.10b), (4.10c), and (4.10d) may be empty. In such a case the constraints in problem (4.10) must be adequately relaxed in order for the problem to be solvable. To this end we rewrite (4.10) as

$$\begin{aligned} & \underset{\tilde{\boldsymbol{\delta}}}{\text{minimize}} \quad f(\tilde{\boldsymbol{\delta}}) \\ & \text{subject to:} \quad \mathbf{A}\tilde{\boldsymbol{\delta}} = \mathbf{b} \\ & \quad \quad \quad \mathbf{C}\tilde{\boldsymbol{\delta}} \leq \mathbf{d} \end{aligned} \quad (4.11)_{\text{new}}$$

By introducing nonnegative slack variables \mathbf{u} , \mathbf{v} , and \mathbf{w} , we relax the problem in (4.11) to

$$\begin{aligned} & \underset{\tilde{\boldsymbol{\delta}}}{\text{minimize}} \quad f(\tilde{\boldsymbol{\delta}}) + \tau \sum_{i=1}^{m+1} (u_i + v_i) + \tau \sum_{j=1}^{2(m+3)} w_j \\ & \text{subject to:} \quad \mathbf{A}\tilde{\boldsymbol{\delta}} - \mathbf{b} = \mathbf{u} - \mathbf{v} \\ & \quad \quad \quad \mathbf{C}\tilde{\boldsymbol{\delta}} - \mathbf{d} \leq \mathbf{w} \\ & \quad \quad \quad \mathbf{u} \geq \mathbf{0}, \mathbf{v} \geq \mathbf{0}, \mathbf{w} \geq \mathbf{0} \end{aligned} \quad (4.12a-d)_{\text{new}}$$

where $\tau > 0$ is a sufficiently large scalar. It is easy to verify that the feasible region defined by (4.12b) – (4.12d) is always nonempty. For example, if we fix $\tilde{\boldsymbol{\delta}} = \tilde{\boldsymbol{\delta}}_0$ arbitrarily, then obviously the point $\{\tilde{\boldsymbol{\delta}}_0, \mathbf{u}_0, \mathbf{v}_0, \mathbf{w}_0\}$ with

$$\mathbf{u}_0 = \max\{\mathbf{0}, A\tilde{\boldsymbol{\delta}}_0 - \mathbf{b}\}, \mathbf{v}_0 = \max\{\mathbf{0}, -(A\tilde{\boldsymbol{\delta}}_0 - \mathbf{b})\}, \text{ and } \mathbf{w}_0 = \max\{\mathbf{0}, C\tilde{\boldsymbol{\delta}}_0 - \mathbf{d}\}$$

Is a feasible point for problem (4.12). The penalty term tries to reduce the magnitudes of the slack variables while minimizing the original objective function $f(\tilde{\boldsymbol{\delta}})$. If the solution slack variables turn out to be all zero, then the solution $\tilde{\boldsymbol{\delta}}$ of (4.12) also solves problem (4.11). Otherwise, we conclude that problem (4.11) is not solvable and the solution obtained by solving (4.12) is a reasonable candidate for the k th iteration to update $\{\mathbf{x}_k, y_k, \mathbf{z}_k\}$.