

## 凸最適化

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$A \in \mathbb{R}^{m \times n}$ ,  $b \in \mathbb{R}^m$ ,  $C \in \mathbb{R}^{n \times n}$  とする. パラメータ  $x = (x_1, \dots, x_n)^\top \in \mathbb{R}^n$  をもつ次の非線形計画問題を考える.

$$\begin{aligned} P(x): \quad & \text{Minimize} \quad \sum_{i=1}^n (z^i)^\top z^i + y^\top y + x^\top C x \\ & \text{subject to} \quad y - \sum_{i=1}^n x_i z^i = Ax - b \end{aligned}$$

ここで,  $P(x)$  の決定変数は  $y, z^i \in \mathbb{R}^m$  ( $i = 1, \dots, n$ ) である. また,  $^\top$  は転置記号を表す. さらに, 任意の  $x$  に対して, 問題  $P(x)$  の最適値が定義されているとし, その最適値を  $f(x)$  と表す.

以下の問い合わせに答えよ.

- (i) 問題  $P(x)$  のカルーシュ・キューン・タッカー条件 (Karush-Kuhn-Tucker 条件) を書け.
- (ii) 問題  $P(x)$  の目的関数が,  $y, z^i \in \mathbb{R}^m$  ( $i = 1, \dots, n$ ) に対して凸であることを示せ.
- (iii)  $C$  を正定値対称行列と仮定し, 次の最適化問題を考える.

$$\begin{aligned} P1: \quad & \text{Minimize} \quad f(x) \\ & \text{subject to} \quad x \in \mathbb{R}^n \end{aligned}$$

$x^* \in \mathbb{R}^n$  を問題  $P1$  の大域的最適解とするとき, 以下の不等式が成り立つことを示せ.

$$(x^*)^\top x^* \leqq \frac{b^\top b}{\lambda_{\min}(C)}$$

ただし,  $\lambda_{\min}(C)$  は  $C$  の最小固有値を表す.

- (iv)  $A$  を  $m \times n$  零行列,  $b$  を  $m$  次元零ベクトルと仮定する. 以下の最適化問題を考える.

$$\begin{aligned} P2: \quad & \text{Minimize} \quad f(x) \\ & \text{subject to} \quad x^\top x \leqq \alpha \end{aligned}$$

ここで,  $\alpha \in \mathbb{R}$  は正の定数である.  $(\hat{x}, \rho), (\bar{x}, \rho) \in \mathbb{R}^n \times \mathbb{R}$  が共に問題  $P2$  のカルーシュ・キューン・タッカー条件を満たすとき,  $f(\hat{x}) = f(\bar{x})$  が成り立つことを示せ.

An English Translation:

## Convex Optimization

### 3

Let  $\mathbf{A} \in \mathbb{R}^{m \times n}$ ,  $\mathbf{b} \in \mathbb{R}^m$  and  $\mathbf{C} \in \mathbb{R}^{n \times n}$ . Consider the following nonlinear programming problem with parameter  $\mathbf{x} = (x_1, \dots, x_n)^\top \in \mathbb{R}^n$ :

$$\begin{aligned} P(\mathbf{x}): \quad & \text{Minimize} \quad \sum_{i=1}^n (\mathbf{z}^i)^\top \mathbf{z}^i + \mathbf{y}^\top \mathbf{y} + \mathbf{x}^\top \mathbf{C} \mathbf{x} \\ & \text{subject to} \quad \mathbf{y} - \sum_{i=1}^n x_i \mathbf{z}^i = \mathbf{A} \mathbf{x} - \mathbf{b}, \end{aligned}$$

where the decision variables of  $P(\mathbf{x})$  are  $\mathbf{y}, \mathbf{z}^i \in \mathbb{R}^m$  ( $i = 1, \dots, n$ ), with  $^\top$  denoting transposition. Moreover, denote by  $f(\mathbf{x})$  the optimal value of problem  $P(\mathbf{x})$ , assuming that it is well-defined for all  $\mathbf{x}$ .

Answer the following questions.

- (i) Write out the Karush-Kuhn-Tucker conditions of  $P(\mathbf{x})$ .
- (ii) Prove that the objective function of problem  $P(\mathbf{x})$  is convex with respect to  $\mathbf{y}, \mathbf{z}^i \in \mathbb{R}^m$  ( $i = 1, \dots, n$ ).
- (iii) Assume that  $\mathbf{C}$  is symmetric positive definite and consider the following optimization problem:

$$\begin{aligned} P1: \quad & \text{Minimize} \quad f(\mathbf{x}) \\ & \text{subject to} \quad \mathbf{x} \in \mathbb{R}^n. \end{aligned}$$

Show that the following inequality holds when  $\mathbf{x}^* \in \mathbb{R}^n$  is a global optimal solution of problem P1:

$$(\mathbf{x}^*)^\top \mathbf{x}^* \leqq \frac{\mathbf{b}^\top \mathbf{b}}{\lambda_{\min}(\mathbf{C})},$$

where  $\lambda_{\min}(\mathbf{C})$  denotes the smallest eigenvalue of  $\mathbf{C}$ .

- (iv) Assume that  $\mathbf{A}$  is the  $m \times n$  zero matrix and  $\mathbf{b}$  is the  $m$ -dimensional zero vector. Consider the following optimization problem:

$$\begin{aligned} P2: \quad & \text{Minimize} \quad f(\mathbf{x}) \\ & \text{subject to} \quad \mathbf{x}^\top \mathbf{x} \leqq \alpha, \end{aligned}$$

where  $\alpha \in \mathbb{R}$  is a positive constant. Show that  $f(\hat{\mathbf{x}}) = f(\bar{\mathbf{x}})$  holds, when both  $(\hat{\mathbf{x}}, \rho), (\bar{\mathbf{x}}, \rho) \in \mathbb{R}^n \times \mathbb{R}$  satisfy the Karush-Kuhn-Tucker conditions of problem P2.