Nonlinear Optimization Exercises 7

1. Given $c \in \mathbb{R}^n$, determine analytically the optimal solution of the problem

min
$$\sum_{i=1}^{n} x_i \ln x_i - c^T x$$
,
s.t. $\sum_{i=1}^{n} x_i = 1$,
 $x > 0$.

2. Prove Theorem IX.8: There is no direction $d \in F_1(\bar{x})$ for which $d^T \nabla f(\bar{x}) < 0$ if and only if there exists a vector $\lambda \in \mathbb{R}^n$ so that

$$\nabla f(\bar{x}) = \sum_{i \in \mathcal{A}(\bar{x})} \lambda_i \nabla c_i(\bar{x}) \text{ with } \lambda_i \ge 0 \text{ for } i \in \mathcal{A}(\bar{x}) \cap \mathcal{I}.$$

Hint: Use Farkas' Lemma.

3. Given the setting of the proof of Theorem IX.13, show that the following matrix is regular,

$$\begin{bmatrix} \nabla^2 f(x^*) - \sum_i \lambda_i^* \nabla^2 c_i(x^*) & -[\nabla c_1(x^*), \dots, \nabla c_m(x^*)] \\ [\nabla c_1(x^*), \dots, \nabla c_m(x^*)]^T & 0 \end{bmatrix}.$$

4. Given a smooth concave function $c: \mathbb{R}^n \to \mathbb{R}$, show that for all $x, y \in \mathbb{R}^n$

$$c(y) \le c(x) + \nabla c(x)^T (y - x).$$

- 5. Given a smooth concave function $c: \mathbb{R}^n \to \mathbb{R}$, show that $-\log(c(x))$ is convex on $\{x \in \mathbb{R}^n : c(x) > 0\}$.
- 6. Consider the following quadratic program in \mathbb{R}^2

$$\begin{aligned} & \min & & \frac{1}{2}x^TQx - bx \\ & \text{s.t.} & & x_1 + 1 \geq 0 \\ & & -x_1 + 1 \geq 0 \\ & & x_2 + 1 \geq 0 \\ & & -x_2 + 1 \geq 0 \end{aligned} .$$

Determine optimal solutions, Lagrange multipiers and (strongly/weakly) active index sets for the cases

(a)
$$Q = \begin{bmatrix} 4 & 1 \\ 1 & 2 \end{bmatrix}$$
, $b = \begin{bmatrix} 14 \\ 7 \end{bmatrix}$ (b) $Q = \begin{bmatrix} 3 & 0 \\ 0 & 2 \end{bmatrix}$, $b = \begin{bmatrix} 6 \\ 1 \end{bmatrix}$

$$(c) \quad Q = \begin{bmatrix} 3 & 0 \\ 0 & 2 \end{bmatrix}, \ b = \begin{bmatrix} 6 \\ 2 \end{bmatrix}$$

What is the infinitesimal change of the value of the optimal solution in each case, if the right hand side is shifted by some small $d \in \mathbb{R}^4$ (use Th.IX.13).