## ELEG5481 Signal Processing Optimization Techniques Assignment 5

Apr. 22, 2013

Deadline: 5:00pm, May. 6, 2013

**Q1.** (30%) Consider the following problem

$$\min_{x} 47x_1 + 93x_2 + 17x_3 - 93x_4 \\ \text{s.t.} \begin{bmatrix} -1 & -6 & 1 & 3\\ -1 & -2 & 7 & 1\\ 0 & 3 & -10 & -1\\ -6 & -11 & -2 & 12\\ 1 & 6 & -1 & -3 \end{bmatrix} \begin{bmatrix} x_1\\ x_2\\ x_3\\ x_4 \end{bmatrix} \preceq \begin{bmatrix} -3\\ 5\\ -8\\ -7\\ 4 \end{bmatrix}.$$

- (a) Write down the dual problem.
- (b) Use CVX to solve the primal and the dual problem to obtain a primal optimal solution  $x^*$  and a dual optimal solution  $\lambda^*$ , respectively. You are required to send via email your matlab code to the tutor.
- (c) Write down the KKT conditions. Then, verify numerically by matlab that  $x^*$  and  $\lambda^*$  satisfy the KKT conditions.

**Q2.** (30%) Implement the subgradient method in matlab to solve the following matrix completion problem.

$$\begin{split} \min_{X} & \|X\|_{*} \\ \text{s.t. } X_{ij} = A_{ij}, \quad (i,j) \in \Omega, \end{split}$$

where  $\{A_{ij}\}_{(i,j)\in\Omega}$  is given. Download test data from http://dsp.ee.cuhk.edu.hk/eleg5481/homework/ hw5\_q2data.mat. The data includes two matrices  $A \in \mathbb{R}^{m \times n}$  and  $I \in \mathbb{R}^{m \times n}$ . I is a zero-one matrix indicating the indices in  $\Omega$ . For example, if  $I_{10,5} = 1$ , then (10,5) belongs to  $\Omega$ ; if  $I_{20,3} = 0$ , then (20,3) does not belong to  $\Omega$ . Write down the 5 × 5 leading principal submatrix of X obtained by your subgradient method. Also, you are required to send via email your matlab code to the tutor.

Hint 1: A subgradient of the nuclear norm  $||X||_*$  is given by  $\sum_{i=1}^r u_i v_i^T$ , where r is the rank of X, and  $u_i$  and  $v_i$  are left and right singular vectors corresponding to positive singular value  $\sigma_r$ .

Hint 2: The projection of X onto the set  $C = \{X \mid X_{ij} = A_{ij}, (i, j) \in \Omega\}$  is as follows. Let  $\tilde{X}$  be the projection of X onto C. We have

$$\tilde{X}_{ij} = \begin{cases} A_{ij}, & (i,j) \in \Omega \\ X_{ij}, & \text{otherwise.} \end{cases}$$

Hint 3: You may use the following parameters: stepsize  $\alpha_k = 10/\sqrt{k}$  and number of iterations 5000. You are encouraged to try other parameters as well.

**Q3.** (20%) Suppose  $a \in \mathbb{R}^n$  with  $a_1 \ge a_2 \ge \ldots \ge a_n > 0$ , and  $b \in \mathbb{R}^n$  with  $b_k = 1/a_k$ . Derive the KKT conditions for the convex optimization problem

$$\min_{x} -\log(a^{T}x) - \log(b^{T}x)$$
  
s.t.  $x \succeq 0$ ,  $\mathbf{1}^{T}x = 1$ .

Show that x = (1/2, 0, ..., 0, 1/2) is optimal.

**Q4.** (a) (10%) Consider the following problem.

$$\min_{X} \operatorname{tr}(CX) \\
\text{s.t. } \operatorname{tr}(AX) \leq 1, \ \operatorname{tr}(BX) = 1, \\
X \succeq 0,$$
(1)

where  $A, B, C \in \mathbf{S}^n$  are symmetric matrices. Assume that this problem is strictly feasible. Show that there exists an optimal solution  $X^*$  whose rank satisfies  $\operatorname{rank}(X^*) \leq 1$ .

Hint: i) You can use the following matrix decomposition result [Ye-Zhang'03]. Suppose that  $X \in \mathbf{S}_{+}^{n}$  has rank r, and  $F \in \mathbf{S}^{n}$  is given symmetric matrix. Then, there exists a set of vectors  $\{x_{1}, \ldots, x_{r}\} \subset \mathbf{R}^{n}$  such that

$$X = \sum_{j=1}^{r} x_j x_j^T$$

and

$$x_j^T F x_j = \frac{\operatorname{tr}(FX)}{r}, \quad j = 1, \dots, r.$$

ii) Use the KKT conditions.

(b) (10%) Consider the S-Lemma, which is widely used in many areas. Let  $F_1, F_2 \in \mathbf{S}^n$ ,  $g_1, g_2 \in \mathbf{R}^n$ , and  $h_1, h_2 \in \mathbf{R}$ . Assume that there exists a point  $x_0$  such that  $x_0F_1x_0 + 2g_1^Tx_0 + h_1 < 0$ . The S-Lemma states that

$$x^T F_2 x + 2g_2^T x + h_2 \leq 0$$
 for all x satisfying  $x^T F_1 x + 2g_1^T x + h_1 \leq 0$ 

holds if and only if there exists a  $\lambda \geq 0$  such that

$$\begin{bmatrix} F_2 & g_2 \\ g_2^T & h_2 \end{bmatrix} \preceq \lambda \begin{bmatrix} F_1 & g_1 \\ g_1^T & h_1 \end{bmatrix}.$$

Prove the S-Lemma by using the result in Q.4(a).

## **Reference:**

Y. Ye and S. Zhang, "New results on quadratic minimization," SIAM J. Optim., 2003.