

THE PRIMAL-DUAL POTENTIAL REDUCTION ALGORITHM FOR POSITIVE SEMI-DEFINITE PROGRAMMING^{*1)}

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Abstract

In this paper we introduce a primal-dual potential reduction algorithm for positive semi-definite programming. Using the symmetric preserving scalings for both primal and dual interior matrices, we can construct an algorithm which is very similar to the primal-dual potential reduction algorithm of Huang and Kortanek [6] for linear programming. The complexity of the algorithm is either $O(n \log(X^0 \bullet S^0/\epsilon))$ or $O(\sqrt{n} \log(X^0 \bullet S^0/\epsilon))$ depends on the value of ρ in the primal-dual potential function, where X^0 and S^0 is the initial interior matrices of the positive semi-definite programming.

Key words: Positive semi-definite programming, Potential reduction algorithms, Complexity.

1. Introduction

In this paper, we consider the following standard form of positive semi-definite programming:

$$(PSP) \quad \text{Minimize} \quad C \bullet X$$

$$\text{Subject to} \quad A_i X = b_i, \quad i = 1, \dots, m, \quad X \succeq 0,$$

where $C, X \in \mathcal{M}^n$, $A_i \in \mathcal{M}^n$, $i = 1, \dots, m$, and $b \in R^m$. Here \mathcal{M}^n denotes the set of symmetric matrices in $R^{n \times n}$. Let \mathcal{M}_+^n denotes the set of positive semi-definite matrices in \mathcal{M}^n and \mathcal{M}_{++}^n denotes the set of positive definite matrices in \mathcal{M}^n . We call \mathcal{M}_{++}^n the interior of \mathcal{M}^n . The notation $X \succeq 0$ means that $X \in \mathcal{M}_+^n$, and $X \succ 0$ means that $X \in \mathcal{M}_{++}^n$. If $X \succ 0$ satisfies all equations in (PSP), it is called a primal interior feasible solution. The \bullet operation is the matrix inner product

$$A \bullet B := \text{tr} A^T B = \sum_{i,j} A_{ij} B_{ij}.$$

The dual problem to (PSP) can be written as:

$$(PSD) \quad \text{Maximize} \quad b^T y$$

$$\text{Subject to} \quad S = C - \sum_{i=1}^m y_i A_i, \quad S \succeq 0,$$

where $S \in \mathcal{M}^n$, $y \in R^m$. If a point $(y, S \succ 0)$ satisfies all equations in (PSD), it is called a dual interior feasible solution.

Define the Frobenius norm, or the l_2 norm, of the matrix $X \in \mathcal{M}^n$ by

$$\|X\| := \|X\|_f = \sqrt{X \bullet X} = \sqrt{\sum_{j=1}^n (\lambda_j(X))^2},$$

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where $\lambda_j(X)$ is the j th eigenvalue of X , and the l_∞ norm of X by

$$\|X\|_\infty := \max_{j \in \{1, \dots, n\}} \{|\lambda_j(X)|\}.$$

Since semi-definite programming has many applications in combinatorial optimization, control theory, statistics, etc., it becomes a hot research topic in optimization over the last decade. Many interior point algorithms have been developed to solve the semi-definite programming. The primal potential reduction algorithms were developed by Alizadeh [1], Nesterov and Nemirovskii [7], Ye [13], etc.; the primal-dual potential reduction algorithms using symmetric matrix scaling were proposed by Nesterov and Todd [8], Kojima, Shindoh and Hara [4], among others. In this paper, we introduce a primal-dual potential reduction algorithm, which uses separate matrices scaling, for above positive semi-definite programming. This kind of scaling has been used extensively in interior point algorithms for linear programming (e.g., Kojima et al. [4], Huang and Kortanek [7],[8], Gonzaga and Todd [6]). To the best of our knowledge we have not seen a paper on interior-point algorithms for semi-definite programming which uses such separate matrices scaling.

To measure the progress of the algorithm, we will use the following primal-dual potential function

$$\phi(X, S) = \rho \log X \bullet S - \log \det XS. \tag{1}$$

The reduction in potential function is controlled by the length of projection of the search directions. In this paper we show that the length of projection is bounded below by $1/4$ if $\rho = n + \sqrt{n}$. Furthermore, we prove that the length is greater than or equal to one if $\rho \geq 2n + \sqrt{2n}$. These results are the extensions of the results in Huang and Kortanek [8] for linear programming to semi-definite programming.

2. The Search Directions

The gradient matrices of (1) are

$$\nabla \phi_X(X, S) = \frac{\rho}{X \bullet S} S - X^{-1}, \tag{2}$$

$$\nabla \phi_S(X, S) = \frac{\rho}{X \bullet S} X - S^{-1}. \tag{3}$$

Let $A = (a_1, \dots, a_n)$ be any $n \times n$ matrix, where a_j ($j = 1, \dots, n$) are columns of A , we define the vector of A as follows:

$$vec(A) = (a_1^T, \dots, a_n^T)^T.$$

Then define

$$\mathcal{A} = \begin{pmatrix} vec(A_1)^T \\ vec(A_2)^T \\ \vdots \\ vec(A_m)^T \end{pmatrix}.$$

Also define the operator $\mathcal{A} : \mathcal{M}^n \rightarrow R^m$ as follows:

$$\mathcal{A}X = A vec(X).$$

Furthermore

$$\mathcal{A}^T y = \sum_{i=1}^m y_i A_i.$$

Given a primal-dual interior feasible solution (X^0, y^0, S^0) such that $\mathcal{A}X^0 = b$ and $S^0 = C - \mathcal{A}^T y^0$, and a $\beta \in (0, 1)$, we consider the following homogeneous minimization problem:

$$\begin{aligned} \text{(HPSD)} \quad & \min \quad \nabla \phi_X(X^0, S^0) \bullet \Delta X + \nabla \phi_S(X^0, S^0) \bullet \Delta S \\ \text{s.t.} \quad & \mathcal{A} \Delta X = 0 \\ & \mathcal{A}^T \Delta y + \Delta S = 0 \\ & \|(X^0)^{-.5} \Delta X (X^0)^{-.5}\|^2 + \|(S^0)^{-.5} \Delta S (S^0)^{-.5}\|^2 \leq \beta^2 < 1. \end{aligned}$$