## On solution of total least squares problems with multiple right-hand sides

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Consider a linear approximation problem  $AX \approx B$  with multiple right-hand sides. When errors in the data are confirmed both to B and A, the total least squares (TLS) concept is used to solve this problem. Contrary to the standard least squares approximation problem, a solution of the TLS problem may not exist. For a single (vector) right-hand side, the classical theory has been developed by G. H. Golub, C. F. Van Loan [2], and S. Van Huffel, J. Vandewalle [4], and then complemented recently by the core problem approach of C. C. Paige, Z. Strakoš [5–7]. Analysis of the problem with multiple right-hand sides is still under development. In this short contribution we present conditions for the existence of a TLS solution.

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## 1 Introduction

Consider an orthogonally invariant linear approximation problem with d right-hand sides

$$AX \approx B, \qquad A \in \mathbb{R}^{m \times n}, \quad X \in \mathbb{R}^{n \times d}, \quad B \in \mathbb{R}^{m \times d},$$
(1)

where the uninteresting case is excluded by the assumption  $A^T B \neq 0$ . With no loss of generality assume  $m \geq n + d$  (add zero rows if necessary). In *total least squares* (TLS) [1,2,4] this problem is solved by constructing minimal correction to B and A such that the corrected system is compatible,

$$\min_{E,G,X} \|[E,G]\|_F \quad \text{subject to} \quad (A+E)X = B+G.$$
(2)

With d = 1 the problem (1) reduces to  $Ax \approx b$ , where b is an m-vector. This problem has been analyzed in [2, 4] and then revised in [5–7], see also [3]. In [5–7], it is shown how the necessary and sufficient information for solving (1) can be extracted from the data using the so called *core reduction*. With the singular value decomposition (SVD)

$$[b, A] = U\Sigma V^T = \sum_{i=1}^{n+1} u_i \sigma_i v_i^T, \qquad \sigma_1 \ge \cdots \ge \sigma_p > \sigma_{p+1} = \cdots = \sigma_{n+1} \ge 0,$$

and the corresponding partitioning

$$V = \begin{bmatrix} v_{11} & v_{12} \\ \hline V_{21} & V_{22} \end{bmatrix}, \quad v_{11} \in \mathbb{R}^{1 \times p}, \, v_{12} \in \mathbb{R}^{1 \times (n-p+1)}, \, V_{21} \in \mathbb{R}^{n \times p}, \, V_{22} \in \mathbb{R}^{n \times (n-p+1)},$$

the classical analysis gives the following conditions on existence and uniqueness of a TLS solution in the case d = 1, see [7] for more details, and the relationship to the core problem theory, which in particular gives an insight into the non-existence of the TLS solution:

- If  $v_{12} \neq 0$  with p = n, i.e.  $\sigma_{n+1}$  is simple and the corresponding right singular vector of [b, A] has a nonzero first component, then there exists the unique TLS solution of (1).
- If  $v_{12} \neq 0$  with p < n, i.e.  $\sigma_{n+1}$  is multiple and there exists right singular vector corresponding to  $\sigma_{n+1}$  with a nonzero first component, then the TLS problem (2) does not have a unique solution. The unique minimum norm solution can be constructed.
- If  $v_{12} = 0$ , i.e. all right singular vectors corresponding to  $\sigma_{n+1}$  have zero first components, then the TLS solution of (1) does not exist.

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## 2 Classification of TLS problems

In order to handle a possible multiplicity of  $\sigma_{n+1}$  in case d > 1, consider the SVD

$$[B, A] = U\Sigma V^T = \sum_{i=1}^{n+d} u_i \sigma_i v_i^T,$$
  
$$\sigma_1 \ge \dots \ge \sigma_p > \sigma_{p+1} = \dots = \sigma_{n+1} = \dots = \sigma_{n+e} > \sigma_{n+e+1} \ge \dots \ge \sigma_{n+d} \ge 0,$$

and the corresponding partitioning

$$V = \begin{bmatrix} V_{11} & V_{12} & V_{13} \\ \hline V_{21} & V_{22} & V_{23} \end{bmatrix}, \qquad V_{11} \in \mathbb{R}^{d \times p}, \ V_{12} \in \mathbb{R}^{d \times (n-p+e)}, V_{13} \in \mathbb{R}^{d \times (d-e)}$$

The matrix  $V_{12}$  represents the first *d* rows of the block of singular vectors of [B, A] corresponding to  $\sigma_{n+1}$ , while  $V_{11}$  and  $V_{13}$  represent the first *d* rows of the block of singular vectors of [B, A] corresponding to singular values strictly greater and strictly smaller than  $\sigma_{n+1}$ , respectively.

The classical analysis [4] of the TLS problem (2) with d > 1 is based on the rank of the matrix  $[V_{12}, V_{13}]$  and deals with three special cases. It is shown that if  $[V_{12}, V_{13}]$  has full row rank d and either p = n (i.e.  $\sigma_n > \sigma_{n+1}$ ) or e = d (i.e.  $\sigma_{n+1} = \sigma_{n+d}$ ), then a TLS solution exists. If rank  $([V_{12}, V_{13}]) < d$  then the TLS solution does not exist, but the TLS concept is in [4] extended to the so called nongeneric solution. The case  $[V_{12}, V_{13}]$  of full row rank d with p < n and e < d, is not analyzed in the literature known to us, and a TLS solution is not defined. The TLS algorithm [4, Algorithm 3.1, pp. 87–88] computes, however, *some* solution X for any problem (1).

The analysis of the TLS problem in [8,9] looks at individual ranks of the matrices  $V_{12}$  and  $V_{13}$ . Let  $[V_{12}, V_{13}]$  be of full row rank d. We can distinguish three cases which offer the classification of TLS problems for d > 1:

- 1. If rank  $([V_{12}, V_{13}]) = d$  with rank  $(V_{12}) = e$  and rank  $(V_{13}) = d e$  (maximal), then there exists a TLS solution of (1). If, moreover, p = n, then the solution is unique.
- **2.** If rank  $([V_{12}, V_{13}]) = d$  with rank  $(V_{12}) > e$  and rank  $(V_{13}) = d e$  (maximal), then there exists a TLS solution of (1).
- **3.** If rank  $([V_{12}, V_{13}]) = d$  with rank  $(V_{12}) > e$  and rank  $(V_{13}) < d e$ , then the TLS solution of (1) does not exist.

It is worth to note that the TLS algorithm [4, Algorithm 3.1, pp. 87–88] computes a TLS solution, i.e. a solution satisfying (2), only in the case 1. The cases  $[V_{12}, V_{13}]$  of full row rank d with p = n or e = d analyzed in [4], are special cases of case 1 in our classification.

This classification may be completed by the fourth case:

4. If rank  $([V_{12}, V_{13}]) < d$ , then the TLS solution of (1) does not exist.

Note that since the classification of situations which can occur is complicated, we can not follow the basic, generic and nongeneric terminology used in [4]. Here we presented several ideas which initiated our work on the subject. Results of a joint effort with Diana M. Sima and Sabine Van Huffel focused on the analysis of TLS problems with multiple right-hand sides, will be presented elsewhere. Extension of the core reduction to problems with multiple right-hand sides is under development.

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