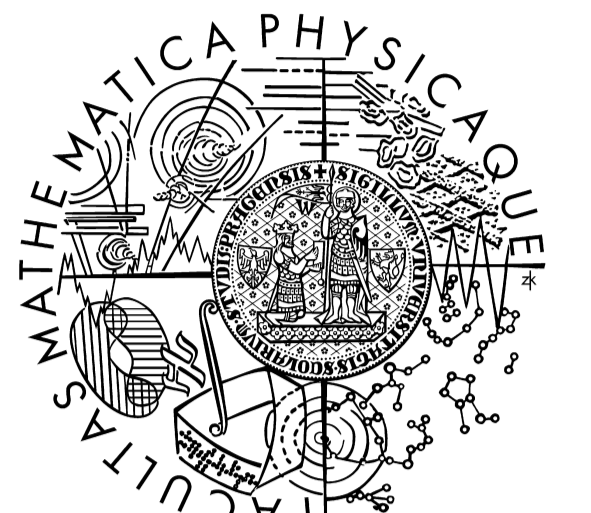
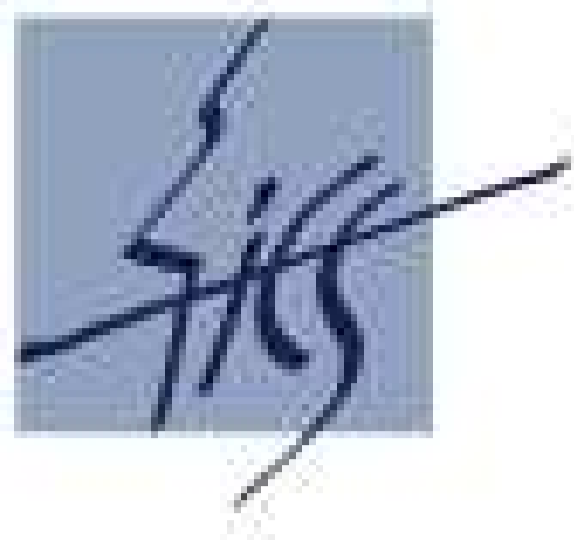


# On total least squares formulation in linear approximation problems with multiple right-hand sides



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## 1. Introduction

Consider an orthogonally invariant linear approximation problem

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

or, equivalently,

$$\begin{bmatrix} B & A \end{bmatrix} \begin{bmatrix} -I_d \\ X \end{bmatrix} \approx 0.$$

With no loss of generality it is assumed:

- $A^T B \neq 0$  (otherwise the columns of  $B$  are not correlated with the columns of  $A$ ,  $X_{\text{TLS}} \equiv 0$ );
- $m \geq n + d$  (add zero rows if necessary).

The linear approximation problem specified by

$$\min_{X, E, G} \| \begin{bmatrix} G & E \end{bmatrix} \|_F \quad \text{s.t.} \quad (A + E)X = B + G \quad (1)$$

is called the **total least squares (TLS) problem** with the TLS solution  $X_{\text{TLS}} \equiv X$  and the correction matrix  $\begin{bmatrix} G & E \end{bmatrix}$ .

When does the TLS solution exist?  
Is it uniquely defined?

## 2. Single right-hand side case

With  $d = 1$ , the problem reduces to  $Ax \approx b$ . Consider the singular value decomposition (SVD)

$$\begin{bmatrix} b & A \end{bmatrix} = U \Sigma V^T$$

with the singular values of  $\begin{bmatrix} b & A \end{bmatrix}$

$$\sigma_1 \geq \dots \geq \sigma_p > \sigma_{p+1} = \dots = \sigma_{n+1} \geq 0,$$

and the partitioning

$$V = \left[ \begin{array}{c|c} \begin{matrix} V_{11} & V_{12} \\ V_{21} & V_{22} \end{matrix} & \begin{matrix} V_{13} \\ V_{23} \end{matrix} \end{array} \right] \begin{matrix} p \\ n-p+1 \end{matrix} \Bigg\} 1 \\ \Bigg\} n$$

(If  $\sigma_1 = \sigma_{n+1}$ , then  $p = 0$  and  $\sigma_p, V_{11}, V_{21}$  are nonexistent.)

Existence and uniqueness of the TLS solution has been analyzed in [1, 4, 5, 6], which gives the following classification:

- If  $V_{12} \neq 0$  with  $p = n$ , then the TLS problem has the **unique (basic) solution**.
- If  $V_{12} \neq 0$  with  $p < n$ , then the TLS problem has infinitely many solutions, the goal is to find the **minimum norm solution**.
- If  $V_{12} = 0$ , then the TLS problem does not have a solution, but the TLS concept can be extended to the so called **nongeneric solution** (that does not solve (1)).

### Core reduction:

In [6] it is shown that for any  $A, b$  there exist orthogonal matrices  $P, Q$  such that

$$P^T \begin{bmatrix} b & A & Q \end{bmatrix} = \begin{bmatrix} b_1 & A_{11} & 0 \\ 0 & 0 & A_{22} \end{bmatrix}.$$

The original problem and its solution are fully defined by two independent subproblems

$$A_{11}x_1 \approx b_1 \quad \text{and} \quad A_{22}x_2 \approx 0,$$

where the first one (called core problem) always has the **TLS solution** and for the second we take  $x_2 \equiv 0$ . Moreover,

$$x \equiv Q \begin{bmatrix} x_1 \\ 0 \end{bmatrix}$$

represents the solution of the original problem identical to one of the corresponding solutions described above (basic, minimum norm, nongeneric).

The core problem concept clarifies the meaning of the **nongeneric solution**.

## 3. Multiple right-hand sides case

Consider the SVD

$$\begin{bmatrix} B & A \end{bmatrix} = U \Sigma V^T$$

with the singular values of  $\begin{bmatrix} B & A \end{bmatrix}$

$$\sigma_1 \geq \dots \geq \sigma_p > \sigma_{p+1} = \dots = \sigma_{n+1} = \dots \\ = \sigma_{n+e} > \sigma_{n+e+1} \geq \dots \geq \sigma_{n+d} \geq 0,$$

and the partitioning

$$V = \left[ \begin{array}{c|c|c} \begin{matrix} V_{11} & V_{12} \\ V_{21} & V_{22} \end{matrix} & \begin{matrix} V_{13} \\ V_{23} \end{matrix} & \begin{matrix} V_{14} \\ V_{24} \end{matrix} \end{array} \right] \begin{matrix} p \\ n+e-p \\ d-e \end{matrix} \Bigg\} d \\ \Bigg\} n$$

(If  $\sigma_1 = \sigma_{n+1}$ , then  $p = 0$  and  $\sigma_p, V_{11}, V_{21}$  are nonexistent. If  $\sigma_{n+1} = \sigma_{n+d}$ , then  $e = d$  and  $\sigma_{n+e+1}, V_{13}, V_{23}$  are nonexistent.)

Classical analysis [4] gives:

- If  $\text{rank}(\begin{bmatrix} V_{12} & V_{13} \end{bmatrix}) = d$  with  $p = n$ , then the TLS problem has the **unique (basic) solution**.
- If  $\text{rank}(\begin{bmatrix} V_{12} & V_{13} \end{bmatrix}) = d$  with  $e = d$ , then the TLS problem has infinitely many solutions, the goal is to find the **minimum norm solution**.
- If  $\text{rank}(\begin{bmatrix} V_{12} & V_{13} \end{bmatrix}) < d$ , then the TLS problem does not have a solution, but the TLS concept can be extended to the so called **nongeneric solution** (that does not solve (1)).

### Difficulty:

The case  $\begin{bmatrix} V_{12} & V_{13} \end{bmatrix}$  of full row rank (equal to  $d$ ) with  $p < n$  and  $e < d$ , is **not analyzed** in the literature, and a TLS solution is not defined. The TLS algorithm [4] by Van Huffel, however computes some  $X$  for any problem  $AX \approx B$ .

Remember that the block  $V_{12}$  corresponds to the singular value  $\sigma_{n+1}$ , while the block  $V_{13}$  corresponds to singular values  $\sigma_j < \sigma_{n+1}$ . In the further analysis we look at *individual ranks* of the matrices  $V_{12}$  and  $V_{13}$ .

### Notation:

When  $\begin{bmatrix} V_{12} & V_{13} \end{bmatrix}$  is of full row rank (equal to  $d$ ), this rank can be “divided” between blocks  $V_{12}$  and  $V_{13}$  in three different ways, see [7].

Denote  $\mathcal{F}_1$  the set of problems (1) for which  $\text{rank}(\begin{bmatrix} V_{12} & V_{13} \end{bmatrix}) = d$  with  $\text{rank}(V_{12}) = e$  and  $\text{rank}(V_{13}) = d - e$  (maximal). (Special cases  $p = n$  and  $e = d$ .)

Denote  $\mathcal{F}_2$  the set of problems (1) for which  $\text{rank}(\begin{bmatrix} V_{12} & V_{13} \end{bmatrix}) = d$  with  $\text{rank}(V_{12}) > e$  and  $\text{rank}(V_{13}) = d - e$  (maximal).

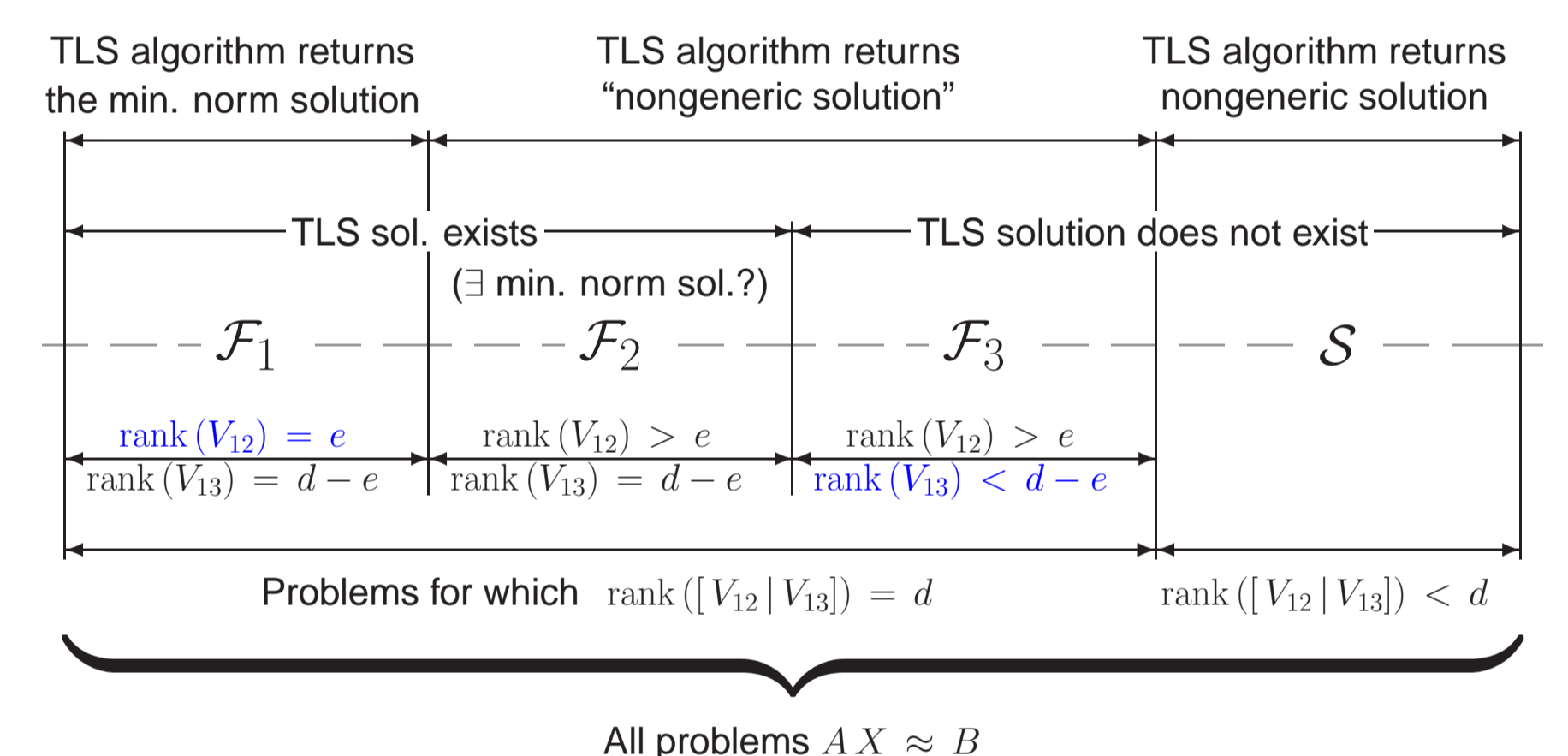
Denote  $\mathcal{F}_3$  the set of problems (1) for which  $\text{rank}(\begin{bmatrix} V_{12} & V_{13} \end{bmatrix}) = d$  with  $\text{rank}(V_{12}) > e$  and  $\text{rank}(V_{13}) < d - e$ .

Denote  $\mathcal{S}$  the set of problems (1) for which  $\text{rank}(\begin{bmatrix} V_{12} & V_{13} \end{bmatrix}) < d$ .

The sets  $\mathcal{F}_j$ ,  $j = 1, 2, 3$ , and  $\mathcal{S}$ , are mutually disjoint and cover all cases that can occur.

## Complete classification of TLS problems, see [7]:

- If the problem belongs in  $\mathcal{F}_1$ , then the TLS problem has a solution. If  $p = n$ , then the solution is unique, otherwise the goal is to find the **minimum norm solution**.
- If the problem belongs in  $\mathcal{F}_2$ , then the TLS problem has a solution, but the **TLS algorithm** [4] does not compute it.
- If the problem belongs in  $\mathcal{F}_3$ , then the TLS problem does not have a solution.
- If the problem belongs in  $\mathcal{S}$ , then the TLS problem does not have a solution, but the TLS concept can be extended to the so called **nongeneric solution** (that does not solve (1)).



## 4. Conclusions

The analysis of the multiple right-hand side problems is complicated. In particular, in the set  $\mathcal{F}_2$  two different “solutions” can be defined – the solution of the minimization problem (1), and the solution given by the TLS algorithm [4]. Extension of the core reduction to multiple right-hand side problems is in progress, see [7].

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