# Splitting of Quasi-Definite Linear System maintains Inertia

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#### Abstract

We show that there is a variety of Schur complements that yield a decoupling of a quasi-definite linear system into two quasi-definite linear systems of half the size each. Splitting of linear systems of equations via Schur complements is widely used to reduce the size of a linear system of equations. Quasi-definite linear systems arise in a variety of computational engineering applications.

### 1 Theorem

**Definitions** Given  $n, m \in \mathbb{N}$ ,  $N = 2 \cdot n$ ,  $M = 2 \cdot m$ ,  $\mathbf{A} \in \mathbb{R}^{N \times N}$ ,  $\mathbf{B} \in \mathbb{R}^{M \times N}$ ,  $\mathbf{D} \in \mathbb{R}^{M \times M}$ , and

$$\hat{\mathbf{K}} := \begin{bmatrix} \mathbf{A} & \mathbf{B}^\mathsf{T} \\ \mathbf{B} & -\mathbf{D} \end{bmatrix},$$

where **A** and **D** are symmetric positive definite.  $\hat{\mathbf{K}}$  is called *quasi-definite* [?].

We consider a splitting of  $\mathbf{A}, \mathbf{B}, \mathbf{D}$  in blocks of size  $n \times n, m \times n$ , and  $m \times m$ , respectively:

$$\mathbf{A} = \begin{bmatrix} \mathbf{A}_1 & \mathbf{A}_2^\mathsf{T} \\ \mathbf{A}_2 & \mathbf{A}_4 \end{bmatrix}, \qquad \mathbf{B} = \begin{bmatrix} \mathbf{B}_1 & \mathbf{B}_2 \\ \mathbf{B}_3 & \mathbf{B}_4 \end{bmatrix}, \qquad \mathbf{D} = \begin{bmatrix} \mathbf{D}_1 & \mathbf{D}_2^\mathsf{T} \\ \mathbf{D}_2 & \mathbf{D}_4 \end{bmatrix}.$$

Using these blocks, we can reorder  $\hat{\mathbf{K}}$  as

$$\mathbf{K} := \begin{bmatrix} \mathbf{A}_1 & \mathbf{B}_1^{\mathsf{T}} & \mathbf{A}_2^{\mathsf{T}} & \mathbf{B}_3^{\mathsf{T}} \\ \hline \mathbf{B}_1 & -\mathbf{D}_1 & \mathbf{B}_2 & -\mathbf{D}_2^{\mathsf{T}} \\ \hline \mathbf{A}_2 & \mathbf{B}_2^{\mathsf{T}} & \mathbf{A}_4 & \mathbf{B}_4^{\mathsf{T}} \\ \hline \mathbf{B}_3 & -\mathbf{D}_2 & \mathbf{B}_4^{\mathsf{T}} & -\mathbf{D}_4 \end{bmatrix} =: \begin{bmatrix} \mathbf{K}_1 & \mathbf{K}_2^{\mathsf{T}} \\ \mathbf{K}_2 & \mathbf{K}_4 \end{bmatrix},$$

where  $\mathbf{K}_1, \mathbf{K}_2, \mathbf{K}_4 \in \mathbb{R}^{(n+m) \times (n+m)}$  are defined as indicated by the 2 × 2 block structure. We then define two matrices:

$$\mathbf{U} := \mathbf{K}_1 - \mathbf{K}_2^{\mathsf{T}} \cdot \mathbf{K}_4^{-1} \cdot \mathbf{K}_2 \tag{1a}$$

$$\mathbf{V} := \mathbf{K}_4 - \mathbf{K}_2 \cdot \mathbf{K}_1^{-1} \cdot \mathbf{K}_2^{\mathsf{T}}$$
(1b)

Notice that the matrices  $\mathbf{U}, \mathbf{V}$  are well-defined since the inverses exist due to quasi-definiteness of  $\mathbf{K}_1, \mathbf{K}_4$ .

**Result** The matrices  $\mathbf{U}, \mathbf{V}$  have the following  $2 \times 2$  block-structure

$$\mathbf{U} = \begin{bmatrix} \mathbf{U}_1 & \mathbf{U}_2^\mathsf{T} \\ \mathbf{U}_2 & -\mathbf{U}_4 \end{bmatrix}, \qquad \mathbf{V} = \begin{bmatrix} \mathbf{V}_1 & \mathbf{V}_2^\mathsf{T} \\ \mathbf{V}_2 & -\mathbf{V}_4 \end{bmatrix},$$

where  $\mathbf{U}_1, \mathbf{V}_1 \in \mathbb{R}^{n \times n}$  and  $\mathbf{U}_4, \mathbf{V}_4 \in \mathbb{R}^{m \times m}$  are symmetric positive definite.

**Proof** All we have to show is that  $U_1$  is symmetric positive definite. The rest then follows by analogy.

Simply by insertion of the blocks into the definition of  $\mathbf{U}$ , we find the following formula for  $\mathbf{U}_1$ :

$$\mathbf{U}_{1} = (\mathbf{A}_{1} + \mathbf{B}_{3}^{\mathsf{T}} \cdot \mathbf{D}_{4}^{-1} \cdot \mathbf{B}_{3}) - (\mathbf{A}_{2} + \mathbf{B}_{4}^{\mathsf{T}} \cdot \mathbf{D}_{4}^{-1} \cdot \mathbf{B}_{3})^{\mathsf{T}} \cdot (\mathbf{A}_{4} + \mathbf{B}_{4}^{\mathsf{T}} \cdot \mathbf{D}_{4}^{-1} \cdot \mathbf{B}_{4})^{-1} \cdot (\mathbf{A}_{2} + \mathbf{B}_{4}^{\mathsf{T}} \cdot \mathbf{D}_{4}^{-1} \cdot \mathbf{B}_{3})$$
(2)

We notice that

$$\mathbf{A}_1 + \mathbf{B}_3^\mathsf{T} \cdot \mathbf{D}_4^{-1} \cdot \mathbf{B}_3$$

is symmetric positive definite because  $\mathbf{A}_1$  and  $\mathbf{D}_4$  are symmetric positive definite.

Hence,  $\mathbf{U}_1$  is positive definite if and only if

$$\mathbf{X} := \begin{bmatrix} \mathbf{A}_1 + \mathbf{B}_3^\mathsf{T} \cdot \mathbf{D}_4^{-1} \cdot \mathbf{B}_3 & (\mathbf{A}_2 + \mathbf{B}_4^\mathsf{T} \cdot \mathbf{D}_4^{-1} \cdot \mathbf{B}_3)^\mathsf{T} \\ \mathbf{A}_2 + \mathbf{B}_4^\mathsf{T} \cdot \mathbf{D}_4^{-1} \cdot \mathbf{B}_3 & \mathbf{A}_4 + \mathbf{B}_4^\mathsf{T} \cdot \mathbf{D}_4^{-1} \cdot \mathbf{B}_4 \end{bmatrix}$$

is positive definite. That is indeed the case, as we now show:

$$\mathbf{x}^{\mathsf{T}} \cdot \mathbf{X} \cdot \mathbf{x}$$
  
=  $\mathbf{x}^{\mathsf{T}} \cdot \mathbf{A} \cdot \mathbf{x} + \left( \begin{bmatrix} \mathbf{B}_3 \\ \mathbf{B}_4 \end{bmatrix} \cdot \mathbf{x} \right)^{\mathsf{T}} \cdot \mathbf{D}_4^{-1} \cdot \left( \begin{bmatrix} \mathbf{B}_3 \\ \mathbf{B}_4 \end{bmatrix} \cdot \mathbf{x} \right) > 0 \qquad \forall \mathbf{x} \in \mathbb{R}^{n+m}, \mathbf{x} \neq \mathbf{0}$ 

because **A** is positive definite and **D** is positive definite, implying that  $D_4$  is positive semi-definite.

**Corollary** From "if and only if" follows that the result holds in a strict sense. I.e.,  $\mathbf{U}_1, \mathbf{V}_1, \mathbf{U}_4, \mathbf{V}_4$  are symmetric positive definite if and only if  $\mathbf{A}, \mathbf{D}$  are symmetric positive definite.

**Remark** The result is easily generalized for complex numbers.

#### 2 Applications

The motivation for considering the matrices  $\mathbf{U}, \mathbf{V}$  is that a linear system of equations with  $\mathbf{K}$  can be decoupled into two linear systems of equations with  $\mathbf{U}, \mathbf{V}$ . We show this.

After suitable reordering of a quasi-definite linear system, consider

$$\mathbf{K} \cdot \begin{bmatrix} \mathbf{x}_1 \\ \mathbf{x}_2 \end{bmatrix} = \begin{bmatrix} \mathbf{b}_1 \\ \mathbf{b}_2 \end{bmatrix} . \tag{3}$$

We find the two decoupled linear systems for  $\mathbf{x}_1, \mathbf{x}_2 \in \mathbb{R}^{n+m}$ :

$$\mathbf{U} \cdot \mathbf{x}_1 = \mathbf{b}_1 - \mathbf{K}_2^{\mathsf{T}} \cdot \mathbf{K}_4^{-1} \cdot \mathbf{b}_2 \tag{4a}$$

$$\mathbf{V} \cdot \mathbf{x}_2 = \mathbf{b}_2 - \mathbf{K}_2 \cdot \mathbf{K}_1^{-1} \cdot \mathbf{b}_1 \tag{4b}$$

Depending on the structure of  $\hat{\mathbf{K}}$ , in some cases it is attractive for computational cost to build and solve the decoupled systems; e.g. when  $\hat{\mathbf{K}}$  is banded [?]. In doing so, one can exploit that the matrices  $\mathbf{U}, \mathbf{V}$  have quasi-definite structure. Also, the result can be applied recursively, as  $\mathbf{U}, \mathbf{V}$  can be considered in turn as matrices of format  $\hat{\mathbf{K}}$ . This recursive applicability can also be used to check the inertia, according to the corollary: The linear systems with  $\mathbf{U}, \mathbf{V}$  are split recursively, until eventually the resulting split systems are so small that one can compute the inertia in a direct way to determine if  $\mathbf{A}, \mathbf{D}$  were positive definite.

Application to Dense Block-Tridiagonal Linear Systems When  $\hat{\mathbf{K}}$  can be reordered into a (dense) block-tridiagonal form, then  $\mathbf{U}, \mathbf{V}$  in turn can be reordered into a dense<sup>1</sup> block-tridiagonal form. This is particularly attractive for solving banded linear systems with a symmetric cyclic reduction algorithm [?] and parallel computations. As we show in [?], formulas for the split linear systems use the matrices  $\mathbf{U}, \mathbf{V}$ . Hence, the theorem holds recursively for all the linear systems generated within the cyclic reduction algorithm, providing following utilities:

- If  $\hat{\mathbf{K}}$  is quasi-definite then by induction so will the matrices  $\mathbf{U}, \mathbf{V}$  in each iteration of the cyclic reduction.
- Following the induction, the inverses in the cyclic reduction algorithm do all exist, since  $\mathbf{K}_1, \mathbf{K}_4$  are quasi-definite (after suitable reordering).
- The inertia of  $\mathbf{U}, \mathbf{V}$  after the last iteration of cyclic reduction can be computed to safeguard to check positive definiteness of  $\mathbf{A}, \mathbf{D}$ .

The last item is particularly useful when the system matrix  $\hat{\mathbf{K}}$  arises within an optimization algorithm. For instance, for SQP methods, which typically compute a step-direction by solving a <u>convex</u> quadratic program [?], the matrix **A** must be ensured to be positive definite, and must be otherwise manipulated (e.g. by applying a shift to **A**).

<sup>&</sup>lt;sup>1</sup>There is fill-in on the band due to the inverses in  $\mathbf{U}, \mathbf{V}$ .

## 3 Conclusion and Outlook

We presented a quite general result for quasi-definite matrices. The result has a wide applicability for structured linear systems, cyclic reduction algorithms, and in the field of convex and non-convex programming.

## References