# 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

$$
{\bf K}:=\left[\begin{array}{cc|cc} {\bf A}_1 & {\bf B}_1^{\sf T} & {\bf A}_2^{\sf T} & {\bf B}_3^{\sf T} \\ {\bf B}_1 & -{\bf D}_1 & {\bf B}_2 & -{\bf D}_2^{\sf T} \\ \hline {\bf A}_2 & {\bf B}_2^{\sf T} & {\bf A}_4 & {\bf B}_4^{\sf T} \\ {\bf B}_3 & -{\bf D}_2 & {\bf B}_4^{\sf T} & -{\bf D}_4 \end{array}\right]=:\left[\begin{array}{cc} {\bf K}_1 & {\bf K}_2^{\sf T} \\ {\bf K}_2 & {\bf K}_4 \end{array}\right]\,,
$$

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

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

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

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

**Result** The matrices **U**, **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 **U**, we find the following formula for **U**1:

$$
\mathbf{U}_1 = (\mathbf{A}_1 + \mathbf{B}_3^T \cdot \mathbf{D}_4^{-1} \cdot \mathbf{B}_3) \n- (\mathbf{A}_2 + \mathbf{B}_4^T \cdot \mathbf{D}_4^{-1} \cdot \mathbf{B}_3)^T \cdot (\mathbf{A}_4 + \mathbf{B}_4^T \cdot \mathbf{D}_4^{-1} \cdot \mathbf{B}_4)^{-1} \n\cdot (\mathbf{A}_2 + \mathbf{B}_4^T \cdot \mathbf{D}_4^{-1} \cdot \mathbf{B}_3)
$$
\n(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  $A_1$  and  $D_4$  are symmetric positive definite.

Hence,  $U_1$  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:

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

because **A** is positive definite and **D** is positive definite, implying that **D**<sup>4</sup> is positive semi-definite.  $\Box$ 

**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 **U***,* **V** is that a linear system of equations with  $\bf{K}$  can be decoupled into two linear systems of equations with **U***,* **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 **K**ˆ is banded [**?**]. In doing so, one can exploit that the matrices **U***,* **V** have quasi-definite structure. Also, the result can be applied recursively, as **U***,* **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 **U***,* **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 **A***,* **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 **U***,* **V** in turn can be reordered into a dense<sup>[1](#page-2-0)</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 **U**, **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 **U***,* **V** after the last iteration of cyclic reduction can be computed to safeguard to check positive definiteness of **A***,* **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 convex 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**).

<span id="page-2-0"></span><sup>1</sup>There is fill-in on the band due to the inverses in **U***,* **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**