



**UNIVERSITÀ DI PISA**

Dipartimento di Informatica  
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# **Computation of Kronecker's Canonical Form in a Computer Algebra System**

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## Introduction

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# Background

This chapter will serve as prerequisite knowledge throughout the rest of this thesis.

We shall briefly present SageMath, the software system used to implement the algorithm discussed in the following chapters, by introducing computer algebra systems and comparing numerical computations against computer algebra showcasing an example; then, the reader shall familiarize with the concept of condition number as an emphasis on it will be put in every part.

Subsequently, definitions and properties of eigenvalues and eigenvectors shall be concisely introduced.

Lastly, we shall describe the Jordan canonical form of a matrix.

## Computer algebra.

Computers have fundamentally two ways to reason about a mathematical expression: **numerical computations**, which are performed using *only numbers* to represent values and **computer algebra** (or **symbolic computations**), which - by contrast - use *both numbers and symbols*.

First, we shall introduce the concept of **floating point number system**, which is the system used to handle numerical computations.

**Definition 2.1** (Normalized-floating point number system). A normalized-floating point number system  $F$  is characterized by the 4-tuple of integers  $\beta, p, L, U$ :

- $\beta$  is called base or radix,
- $p$  precision,
- $[L, U]$  exponent range (with  $L, U$  denoting lower and upper bound respectively).

Given a number  $x \in \mathbb{R}$ ,  $x \neq 0$  its representation in a floating point number system shall be written out as  $fl(x)$  and has the form

$$x = \text{sign}(x)\beta^E \sum_{i=0}^{p-1} d_i \beta^{-i}$$

with  $L \leq E \leq U$  and the sequence  $\{d_i\}_{i \geq 0}$  (which is called mantissa) made up of natural numbers such that  $d_0 \neq 0$ ,  $0 \leq d_i \leq \beta - 1$  and  $d_i$  eventually different from  $\beta - 1$ .

The notation  $\delta x$  shall be used to denote the difference between a symbol  $x$  and its floating point approximation  $fl(x)$

$$\delta x = x - fl(x).$$

It is important to notice that a floating point number system  $F$  is discrete and finite: it approximates real numbers with finite numbers; in other words, a floating point number system may introduce errors when representing a real number.

A de facto standard for computers to work with floating point approximations is IEEE 754 [7], the details of which shall not be discussed.

**Definition 2.2** (Machine epsilon). Machine epsilon is the maximum possible absolute relative error in representing a nonzero real number  $x$  in a floating point number system

$$\epsilon_{mach} = \max_x \frac{|x - fl(x)|}{|x|}.$$

**Example 2.1.** Let us define the matrix (made up of both symbols and numbers)  $M$

$$\begin{bmatrix} \sqrt{2} & 1 \\ 2 & \sqrt{2} \end{bmatrix}.$$

Consider the matrix  $\tilde{M}$ , having as entries the floating point approximation of those of  $M$

$$\begin{bmatrix} fl(\sqrt{2}) & 1 \\ 2 & fl(\sqrt{2}) \end{bmatrix}.$$

Computing its determinant gives out  $2 + 2\epsilon\sqrt{2} + \epsilon^2 - 2 \doteq 2 + 2\epsilon\sqrt{2} - 2 \neq 0$ .

Introducing a small change (i.e. an “error”) in the input argument may either cause a large or a small change in the result. Now, we shall define what condition numbers are.

**Definition 2.3** (Condition number). A condition number of a problem measures the sensitivity of the solution to small perturbations in the input data. Given a function  $f$ , we define

$$cond(f, x) = \lim_{\epsilon \rightarrow 0} \sup_{\|\Delta x\| \leq \epsilon \|x\|} \frac{\|f(x + \Delta x) - f(x)\|}{\epsilon \|f(x)\|}.$$



Given a problem, if its condition number is low it is said to be **well-conditioned** (typically  $\text{cond}(f, x) \sim 1$ ), while a problem with a high condition number is (said to be) **ill-conditioned** ( $\text{cond}(f, x) \gg 1$ ).

Let us now consider the problem of solving a linear equation subjected to a perturbation.

Let  $A$  be a non-singular matrix and assume we introduce a perturbation in the constant term  $\tilde{\mathbf{b}} = \mathbf{b} + \delta\mathbf{b}$ . The equation can be written as

$$A\tilde{\mathbf{x}} = \tilde{\mathbf{b}}$$

with  $\tilde{\mathbf{x}} = \mathbf{x} + \delta\mathbf{x}$ . We can obtain

$$\tilde{\mathbf{x}} - \mathbf{x} = A^{-1}\tilde{\mathbf{b}} - A^{-1}\mathbf{b} = A^{-1}\delta\mathbf{b}$$

and, by using matrix norms, we can write

$$\|\tilde{\mathbf{x}} - \mathbf{x}\| = \|A^{-1}\delta\mathbf{b}\| \leq \|A^{-1}\| \|\delta\mathbf{b}\|.$$

It is also known that

$$\|\mathbf{b}\| = \|A\mathbf{x}\| \leq \|A\| \|\mathbf{x}\|$$

which implies

$$\|\mathbf{x}\| \geq \frac{\|\mathbf{b}\|}{\|A\|}.$$

Tying all this together we can conclude

$$\frac{\|\tilde{\mathbf{x}} - \mathbf{x}\|}{\|\mathbf{x}\|} \leq \|A\| \|A^{-1}\| \frac{\|\delta\mathbf{b}\|}{\|\mathbf{b}\|}.$$

**Definition 2.4** (Condition number of a matrix). The condition number of a non-singular matrix  $A$  is defined as:

$$\kappa(A) = \|A^{-1}\| \|A\|.$$

Now, let us refocus on the topic of math expressions. Let us investigate what would happen if symbols are allowed in computations by introducing a framework that allows us to work with computer algebra.

**Definition 2.5** (Computer algebra system). A computer algebra system (CAS) is a mathematics software package that can perform *both symbolic and numerical mathematical computations*.

A CAS is usually a **REPL** expected to support a few functionalities [8]:

- **Arithmetic:** arithmetic over different fields with arbitrary precision.
- **Linear algebra:** matrix algebra and knowledge of different operations and properties of matrices (i.e. determinants, eigenvalues and eigenvectors).
- **Polynomial manipulation:** factorization over different fields, simplification and partial fraction decomposition of rational functions.
- **Transcendental functions:** support for transcendental functions and their properties.
- **Calculus:** limits, derivatives, integration and expansions of functions.
- **Solving equations:** solving systems of linear equations, computing with radicals solutions of polynomials of degree less than five.
- **Programming language:** users may implement their own algorithms using a programming language.

The CAS chosen for this work is **SageMath** [11], the features and functionalities of which shall not be discussed here.

SageMath is an open source CAS distributed under the terms of the GNU GPLv3 [6].

Hereafter, an example in which symbolic computations are put against numerical (computations) shall be made.

**Example 2.2.** Take matrix M from Example 2.1:

$$\begin{bmatrix} \sqrt{2} & 1 \\ 2 & \sqrt{2} \end{bmatrix}.$$

Compare the different results given out when computing its determinant by defining M over the *symbolic ring SR* and the *finite-precision ring CDF*:

---

```
sage: matrix(SR, [[sqrt(2), 1], [2, sqrt(2))]).det()
0
sage: matrix(CDF, [[sqrt(2), 1], [2, sqrt(2))]).det()
-3.14018491736755e-16
```

---

We can observe that in SR  $(\sqrt{2})^2 = 2$  since no approximations are made.

Now, take the polynomial  $p(x)$ :

$$p(x) = x^6 + 5x^5 - 3x^4 - 42x^3 + 12x^2 - x + 1.$$

If an attempt to calculate its roots over SR is made an exception will be thrown (here, a reader may refer to Abel-Ruffini theorem for further explanations); however, doing this over a finite-precision ring (such as CDF) will work:

---

```
sage: p = x^6 + 5*x^5 - 3*x^4 - 42*x^3 + 12*x^2 - x + 1
sage: p.roots(ring=SR)
      RuntimeError: no explicit roots found
sage: p.roots(ring=CDF)
[(-3.865705050148171 - 1.5654017866113432*I, 1),
 (-3.8657050501481702 + 1.5654017866113419*I, 1),
 (-0.04843174828928114 - 0.2430512799158686*I, 1),
 (-0.048431748289281144 + 0.24305127991586856*I, 1),
 (0.38275295887213723 + 7.286537374692244e-17*I, 1),
 (2.4455206380027437 - 1.995314986816126e-16*I, 1)]
```

---

What we may conclude from such an example is that numerical analysis is certainly a powerful tool as it allows for computations which could not happen with computer algebra, **but** computer algebra being able to compute an exact answer without any approximation will prove to be useful in our use case.

For deeper reasoning about the limits of computer algebra systems, one may refer to Mitic [10].

## Eigenvalues, eigenvectors

In the following section, we shall define **eigenvalues** and **eigenvectors** and discuss the numerical stability of their computation; a reader may also consult Axler [1] or Strang [12] for further explanations.

**Definition 2.6** (Eigenvalue, eigenvector). Given a linear transformation  $T$  in a finite-dimensional vector space  $V$  over a field  $F$  into itself and a nonzero vector  $\mathbf{v}$ ,  $\mathbf{v}$  is an eigenvector of  $T$  if and only if

$$A\mathbf{u} = \lambda\mathbf{u}$$

with  $A$  the matrix representation of  $T$ ,  $\mathbf{u}$  the coordinate vector of  $\mathbf{u}$  and  $\lambda$  a scalar in  $F$  known as eigenvalue associated with  $\mathbf{v}$ .

Similarly, we can define a row vector  $\mathbf{x}_L$ , and a scalar  $\lambda_L$  such that

$$\mathbf{x}_L A = \lambda_L \mathbf{x}_L,$$

which are called **left eigenvector** and **left eigenvalue** respectively.

**Remark.** Note that writing  $A\mathbf{u} = \lambda\mathbf{u}$  is equivalent to  $(A - \lambda I)\mathbf{u} = 0$ .

It follows that the eigenvalues of  $A$  are the roots of

$$\det(A - \lambda I)$$

which is a polynomial in  $\lambda$  known as the **characteristic polynomial**  $ch_A(\lambda)$ .

**Definition 2.7** (Eigenspace). Given a square matrix  $A$  and its eigenvalue  $\lambda$ , we define the eigenspace of  $A$  associated with  $\lambda$  the subspace  $E_A$  of all vectors satisfying the equation

$$E_A = \{\mathbf{u} : (A - \lambda I)\mathbf{u} = 0\} = \ker(A - \lambda I).$$

**Definition 2.8** (Algebraic, geometric multiplicities of eigenvalues). Given a square matrix  $A$  and a scalar  $\lambda \in \mathbb{C}$ : we define the algebraic multiplicity of  $\lambda$  as

$$m_A(\lambda) = \max\{k : (\exists s(x) : s(x)(x - \lambda)^k = ch_A(x))\}.$$

The geometric multiplicity of  $\lambda$  is defined as

$$\nu_A(\lambda) = \dim(\ker(A - \lambda I)).$$

**Remark.** Suppose  $A$  is a real square matrix, then the following statements are true:

- the eigenvalues of the left and right eigenvectors of  $A$  are the same,
- the left eigenvectors simplify into the transpose of the right eigenvectors of  $A^T$ .

Now, let us investigate how introducing perturbations in the representation of a matrix may influence the numerical stability of its eigenvalues.

Let  $A$  be a square matrix,  $\lambda \in \mathbb{C}$  its eigenvalue,  $\mathbf{x}$ ,  $\mathbf{y}$  the right and left eigenvectors associated with  $\lambda$ . Consider the perturbed problem

$$\tilde{A}\tilde{\mathbf{x}} = \tilde{\lambda}\tilde{\mathbf{x}}$$

with  $\epsilon$  the machine epsilon,  $\tilde{A} = A + \epsilon\delta A$ ,  $\tilde{\mathbf{x}} = \mathbf{x} + \epsilon\delta\mathbf{x}$ ,  $\tilde{\lambda} = \lambda + \epsilon\delta\lambda$ .

Differentiating w.r.t.  $\epsilon$  and multiplying by  $\mathbf{y}^T$  on the left side gives

$$\mathbf{y}^T \delta A \mathbf{x} + \mathbf{y}^T A f l(\mathbf{x}) = f l(\lambda) \mathbf{y}^T \mathbf{x} + \mathbf{y}^T \lambda f l(\mathbf{x})$$

and, since  $\mathbf{y}$  is the left eigenvector we can rewrite it as

$$\frac{\delta \lambda}{\delta \epsilon} = \frac{\mathbf{y}^T \delta A \mathbf{x}}{\mathbf{y}^T \mathbf{x}}.$$

Assuming  $\|\delta A\| = 1$  and using the definition of dot product for  $\mathbf{y}^T \mathbf{x}$  we get

$$|\delta \lambda| \leq \frac{1}{|\cos(\theta_\lambda)|} |\delta \epsilon|.$$

**Definition 2.9** (Condition number of an eigenvalue). Given a square matrix  $A$ , the eigenvalue  $\lambda \in \mathbb{C}$  and  $\theta_\lambda$  the angle between the left and right eigenvectors associated with  $\lambda$ , the quantity

$$k_A(\lambda) = \frac{1}{\cos(\theta_\lambda)}$$

is called the condition number of the eigenvalue  $\lambda$ .

## Jordan canonical form

In the following section, we shall define **Jordan matrices** and discuss the stability of a transformation of a matrix into its Jordan canonical form.

**Definition 2.10** (Jordan matrix). A diagonal block matrix  $M$  is called a Jordan matrix if and only if each block along the diagonal is of the form

$$\begin{bmatrix} \lambda & 1 & 0 & \cdots & 0 \\ 0 & \lambda & 1 & \cdots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \lambda & 1 \\ 0 & 0 & 0 & 0 & \lambda \end{bmatrix},$$

and we can write such a matrix  $M$  as  $M = \text{diag}(J_{\lambda_1, n_1}, \dots, J_{\lambda_k, n_k})$  with  $k$  the number of diagonal blocks it is made up of.

Each  $n \times n$  block can be completely characterized by the tuple  $(\lambda, n)$  as it can fully describe both the structure and the dimension of a block.

**Remark.** Let  $V$  be a vector space defined over a field  $F$  and  $A$  a matrix defined in  $V$ . If the characteristic polynomial of  $A$   $ch_A(t)$  can be factorized into its linear factors over  $K$ , then  $A$  is similar to a Jordan matrix  $J$ . We define  $J$  the **Jordan canonical form (JCF)** of  $A$ .

**Definition 2.11** (Defective matrix, defective eigenvalue). Given a square  $n \times n$  matrix  $A$ , if it has less than  $n$  distinct eigenvalues then it is called a defective matrix.

Furthermore, we define an eigenvalue  $\lambda$  of such a matrix as a defective eigenvalue if and only if

$$m_A(\lambda) > \nu_A(\lambda).$$

Now, we shall give a result on the stability of such a transformation the proof of which can be found in other works, such as Datta [3].

**Theorem 2.1** (Stability of the JCF transformation). Given a matrix  $A$  and its JCF  $A = P^{-1}JP$ , the transforming matrix  $P$  is highly ill-conditioned whenever  $A$  has at least a defective or nearly defective eigenvalue.

Lastly, we shall give an example to show the implications of this theorem by showing the differences in the JCF of a matrix and its perturbed version.

**Example 2.3.** Consider the  $n \times n$  matrices

$$A = \begin{bmatrix} 0 & 1 & 0 & \cdots & 0 \\ 0 & 0 & 1 & \cdots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}, B = \begin{bmatrix} 0 & 1 & 0 & \cdots & 0 \\ 0 & 0 & 1 & \cdots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & 0 & 1 \\ \alpha & 0 & 0 & 0 & 0 \end{bmatrix}$$

with  $\alpha > 0$ .

It is evident how  $A$  has a defective eigenvalue in  $\lambda_A = 0$  and  $m_A(0) = n$ ,  $\nu_A(0) = 1$ ; furthermore,  $A$  is already in JCF.

Now, let us switch our focus to  $B$ . To compute its eigenvalues, take the characteristic polynomial  $ch_B(t) = t^n - \alpha$ : it has  $n$  distinct roots in

$$t_j = z_n^j \sqrt[n]{\alpha}$$

with  $j = 1, \dots, n$ ,  $z_n = \cos\left(\frac{2\pi}{n}\right) + i \sin\left(\frac{2\pi}{n}\right)$  and  $i$  imaginary unit such that  $i^2 = -1$ .

To conclude, we shall show the JCF of  $A$  and  $B$  defined in SR computed by SageMath when  $n = 4$ .

---

```

sage: A = matrix(SR, [
    [1 if i == j-1
     else 0 for j in range(4)]
    for i in range(4)
])
sage: B = matrix(SR, [
    [1 if i == j-1
     else x if j == 0 and i == 3
     else 0 for j in range(4)]
    for i in range(4)
])
sage: A
[0 1 0 0]
[0 0 1 0]
[0 0 0 1]
[0 0 0 0]
sage: A.jordan_form()
[0 1 0 0]
[0 0 1 0]
[0 0 0 1]
[0 0 0 0]
sage: B
[0 1 0 0]
[0 0 1 0]
[0 0 0 1]
[x 0 0 0]
sage: B.jordan_form()
[ I*x^(1/4) | 0 | 0 | 0]
[-----+-----+-----+-----]
[ 0 | -x^(1/4) | 0 | 0]
[-----+-----+-----+-----]
[ 0 | 0 | -I*x^(1/4) | 0]
[-----+-----+-----+-----]
[ 0 | 0 | 0 | x^(1/4)]

```

---

For the sake of clarity, we shall also show what the implications of  $B$

having such eigenvalues are.

Suppose  $x = 10^{-10}$ .

---

```
sage: B = matrix(SR, [
    [1 if i == j-1
      else 10**(-10) if j == 0 and i == 3
      else 0 for j in range(4)]
    for i in range(4)
  ])
sage: B
[
  0      1      0      0]
[
  0      0      1      0]
[
  0      0      0      1]
[1/10000000000  0      0      0]
sage: P = B.jordan_form(transformation=True)[1]
sage: cond = norm(P.inverse()) * norm(P)
sage: cond
31622776.60168379
```

---

We can see that  $\kappa(P) \gg 1$ , as stated in [Stability of the JCF transformation](#) (theorem 2.1).



# Theory and applications of pencils of matrices

This chapter will introduce the reader to the concept of a linear pencil of matrices and its properties. Throughout this chapter and the following one, a reader may refer to Gantmacher [4], Kunkel, Mehrmann [9] and Beelen, Van Dooren [2].

**Definition 3.1** (Linear matrix pencil). A linear pencil of matrices is defined as a 1-degree polynomial with matrix coefficients

$$\Gamma(\lambda) = A + \lambda B$$

with  $\lambda \in \mathbb{C}$ ,  $A$  and  $B$   $m \times n$  matrices. A linear pencil of matrices may also be called a **pair of matrices** and, in this thesis, we shall use synonymously both terms.

## Regular pencils.

Now, we consider the case where  $(A + \lambda B)$  is a regular pencil of matrices.

**Definition 3.2** (Regular pencil). A matrix pair  $(A, B)$  is said to be regular if and only if  $A$  and  $B$  are square matrices of the same size and the determinant  $\det(A + \lambda B)$  is not identically zero.

Consider the regular pencil of matrices

$$\Gamma(\lambda) = A + \lambda B,$$

let  $F$  be the field the entries of  $A$  and  $B$  belong to and  $r$  the rank of the pencil.

Denote with  $D_j(\lambda)$  the greatest common divisor of all minors of order  $j$  of  $\Gamma(\lambda)$  (with  $j = 1, \dots, r$ ) and assume without any loss of generality  $D_j(\lambda)$  is monic and  $D_0(\lambda) = 1$ . Given the sequence,

$$D_r(\lambda), D_{r-1}(\lambda), \dots, D_1(\lambda), D_0(\lambda)$$

we define the **invariant polynomials** of the pencil of matrices  $\Gamma(\lambda)$  as the fractions

$$i_1(\lambda) = \frac{D_r(\lambda)}{D_{r-1}(\lambda)}, i_2(\lambda) = \frac{D_{r-1}(\lambda)}{D_{r-2}(\lambda)}, \dots, i_r(\lambda) = D_1(\lambda).$$

We can now write the expansion of the invariant polynomials into irreducible factors in  $F$  as

$$i_1(\lambda) = \prod_{i=1}^k p_i(\lambda)^{\alpha_{1,i}}, \quad i_2(\lambda) = \prod_{i=1}^k p_i(\lambda)^{\alpha_{2,i}}, \quad \dots$$

$$i_r(\lambda) = \prod_{i=1}^k p_i(\lambda)^{\alpha_{r,i}},$$

with  $p_i$  an irreducible polynomial appearing in the expansion.

We define the **elementary(finite) divisors**  $e_i$  of the pencil of matrices  $\Gamma(\lambda)$  all the polynomials  $p_i(\lambda)^{\alpha_{j,i}}$  (with  $j = 1, \dots, r$ ) that are not equal to one.

A similar procedure may be defined for the pencil of matrices

$$\Theta(\mu, \lambda) = \mu A + \lambda B$$

leading to polynomials in two variables  $(\mu, \lambda)$ . Clearly, having  $\mu = 1$  would lead to obtaining the elementary finite divisors of  $\Gamma(\lambda)$ ; however, for each elementary divisor of degree  $q$  we have

$$e_i(\mu, \lambda) = \mu^q e_i\left(\frac{\lambda}{\mu}\right),$$

and, with this technique, it is possible to generate all the elementary divisors of  $\Theta(\mu, \lambda)$  except for those of the form  $\mu^q$ , which are called **elementary infinite divisors** of the pencil of matrices  $\Theta(\mu, \lambda)$ .

**Remark.** A regular pencil of matrices  $\Gamma(\lambda) = A + \lambda B$  has elementary infinite divisors if and only if  $\det(B) = 0$ .

We can now give a result on the equivalence of regular pencils.

**Theorem 3.1** (Equivalence of regular pencils of matrices). Two regular matrix pairs

$$\Gamma_1(\lambda) = (A_1, B_1) \qquad \Gamma_2(\lambda) = (A_2, B_2)$$

are equivalent if and only if they have the same finite and infinite elementary divisors.

In other words,  $\Gamma_1$  and  $\Gamma_2$  are equivalent if and only if the invariant

polynomials of the two pencils  $\Theta_1(\mu, \lambda)$  and  $\Theta_2(\mu, \lambda)$  are equivalent, with

$$\Theta_1(\mu, \lambda) = \mu A + \lambda B \quad \Theta_2(\mu, \lambda) = \mu A + \lambda B$$

**Lemma 3.2.** For any two regular equivalent pencils of  $n \times n$  matrices the entries of which lie in a field  $F$

$$\Gamma_1(\lambda) = A_1 + \lambda B_1 \quad \Gamma_2(\lambda) = A_2 + \lambda B_2$$

there exist two nonsingular matrices  $P \in F^{n \times n}$ ,  $Q \in F^{n \times n}$  such that

$$P\Gamma_1(\lambda)Q = \Gamma_2(\lambda),$$

and we call such a transformation an **equivalence transformation**.

### Singular pencils.

Next, we shall investigate the most general case of  $m \times n$  pencils of matrices in order to introduce the reader to the concept of minimal indices of a pencil of matrices.

**Definition 3.3** (Singular pencil). A matrix pair  $A, B$  is said to be singular if and only if it is not regular.

Consider the singular pencil of (rectangular) matrices  $\Gamma(\lambda) = A + \lambda B$  and assume its rank  $r$  is smaller than its number of columns  $n$ .

This implies the equation

$$(A + \lambda B)\mathbf{x} = 0$$

has nontrivial solutions  $\mathbf{x}_1(\lambda), \dots, \mathbf{x}_k(\lambda)$  and let  $X$  be the polynomial matrix made up of such polynomials

$$X = \begin{bmatrix} x_{1,1} & x_{2,1} & \dots & x_{k,1} \\ x_{1,2} & x_{2,2} & \dots & x_{k,2} \\ \vdots & & & \vdots \\ x_{1,n} & x_{2,n} & \dots & x_{k,n} \end{bmatrix}$$

The columns  $\mathbf{x}_i(\lambda)$  (with  $i = 1, \dots, k$ ) of  $X$  can be chosen to be linearly independent; as a matter of fact, the columns are linearly dependent if the rank of  $X$  is less than  $k$  and, (only) in this case, we can choose  $k$  nontrivial

polynomials  $p_i(\lambda)$  such that

$$\sum_{i=1}^k p_i(\lambda) \mathbf{x}_i(\lambda) \equiv 0.$$

We choose the nontrivial polynomial  $\mathbf{x}_1(\lambda)$  of least degree  $\epsilon_1$  in  $\lambda$ ; next, amongst the very same solutions, we choose  $\mathbf{x}_2(\lambda)$  of least degree  $\epsilon_2$  etc.

Since the maximum number of linearly independent solutions of the aforementioned equation is no more than  $n - r$ , this process has a finite number of steps.

To summarize, from an equation  $(A + \lambda B)\mathbf{x} = 0$  we can obtain a sequence of solutions of non-increasing degree

$$\epsilon_1 \leq \epsilon_2 \leq \dots \leq \epsilon_p,$$

and we define  $\epsilon_i$  a **minimal index for the columns** of the pencil of matrices  $\Gamma(\lambda)$ .

We can also introduce **minimal indices for the rows**  $\eta_1, \eta_2, \dots, \eta_q$  of the pencil  $(A + \lambda B)$  which we can yield working with the transpose of the pencil  $A^T + \lambda B^T$  or, in other words, with the equation

$$(A^T + \lambda B^T)\mathbf{y} = 0.$$

We can now give another result on the equivalence of pencils of matrices.

**Theorem 3.3** (Necessary condition for the equivalence of pencils). Two arbitrary equivalent pencils have the same minimal indices for rows and columns.

## Kronecker canonical form.

At this point, we can put together the theorems on the equivalence of regular and singular pencil of matrices, namely [Equivalence of regular pencils of matrices](#) (theorem 3.1) and [Necessary condition for the equivalence of pencils](#) (theorem 3.3) and give a result on the **equivalence of arbitrary matrix pairs**.

**Theorem 3.4** (Kronecker). Two pencils  $(A + \lambda B)$ ,  $(A_1 + \lambda B_1)$  of rectangular  $m \times n$  matrices are equivalent if and only if they have the same minimal indices for rows and columns and the same elementary finite and infinite divisors.

**Remark.** Restating [Kronecker](#) (theorem 3.4) using different words, we can say that a matrix pair  $(A, B)$  is completely characterized by its minimal indices for rows and columns and (its) elementary finite and infinite divisors and does not depend on their order.

Coherently, it should be possible to define a canonical form for pencils of matrices completely determined by both the minimal indices for rows and columns and the elementary finite and infinite divisors, and it is. We can now introduce the Kronecker canonical form of a matrix pair.

**Theorem 3.5** (Kronecker canonical form). Let  $\Gamma(\lambda) = A + \lambda B$  be an arbitrary pencil of matrices and  $h, g$  the maximal number of constant independent solutions of the two equations

$$(A + \lambda B)\mathbf{x} = 0 \quad (A^T + \lambda B^T)\mathbf{y} = 0.$$

The pencil  $\Gamma(\lambda)$  is strictly equivalent to a quasi-diagonal matrix

$$\{Z; L_{\epsilon_{g+1}}, L_{\epsilon_{g+2}}, \dots, L_{\epsilon_p}; L_{\eta_{h+1}}^T, L_{\eta_{h+2}}^T, \dots, L_{\eta_q}^T; N^{(u_1)}, N^{(u_2)}, \dots, N^{(u_s)}; J + \lambda I\},$$

where  $Z$  is an  $h \times g$  null matrix,  $L_\epsilon$  a rectangular  $\epsilon \times (\epsilon + 1)$  matrix of the form

$$L_\epsilon = \begin{bmatrix} \lambda & 1 & 0 & \dots & 0 & 0 \\ 0 & \lambda & 1 & & \vdots & \vdots \\ \vdots & \vdots & & \ddots & & \\ 0 & 0 & & & \lambda & 1 \end{bmatrix}$$

and

$$N^u = I^u + \lambda H^u$$

with  $I$  the  $u \times u$  identity matrix and  $H$  the  $u \times u$  upper shift matrix, and  $J$  a Jordan matrix.

We shall call such a matrix the Kronecker canonical form of  $\Gamma(\lambda)$  and write it out as  $K(\Gamma(\lambda))$ .

For a proof of [Kronecker](#) (theorem 3.4) and [Kronecker canonical form](#) (theorem 3.5) the reader shall refer to the following chapter.

**Corollary 3.5.1.** Given a matrix pair  $(A, B)$  there exists a pair of nonsingular matrices  $P, Q$  such that

$$PAQ = K(A) \quad PBQ = K(B),$$

with  $P$  an  $m \times m$  and  $Q$  an  $n \times n$  matrices respectively.

**Example 3.1.** Kronecker's canonical form of the pencil of matrices

$$\begin{bmatrix} \boxed{0} & & & & & \\ & \boxed{\lambda} & & & & \\ & \boxed{1} & & & & \\ & & \boxed{1} & & & \\ & & & \boxed{\begin{matrix} 1 & \lambda \\ 0 & 1 \end{matrix}} & & \\ & & & & \boxed{\lambda + 1} & \end{bmatrix}.$$

is made up of three rectangular blocks  $Z^{1 \times 1}$ ,  $L_0$ ,  $L_1^T$ , two nilpotent blocks  $N^{(1)}$ ,  $N^{(2)}$  and one Jordan block  $J + \lambda I^{(1)}$ .

## Fundamental applications.

The following section shall introduce the reader to differential-algebraic equations (which we may shorten as DAE) in order to explain what is the use case of the KCF of a pencil of matrices while keeping a strong focus on examples.

**Definition 3.4** (Differential-algebraic equation). An equation

$$F(t, x, \dot{x}) = 0$$

with  $F : I \times D_x \times D_x \rightarrow \mathbb{C}^m$ ,  $I \subseteq \mathbb{R}$  is a compact interval,  $D_x, D_x \subseteq \mathbb{C}^n$  are open,  $x : I \rightarrow \mathbb{C}^n$  a differentiable function and  $m, n \in \mathbb{N}$  is called differential-algebraic equation.

What we're attempting to determine is a function  $x$  such that

$$\forall x \in I. F(t, x(t), \dot{x}(t)) = 0.$$

Differential-algebraic equations are used to model physical systems with a dynamic behaviour the states of which are subjected to certain constraints. For our purposes, we'll narrow the definition given above and rewrite it as follows.

**Definition 3.5** (Linear DAE with constant coefficients). An equation

$$B\dot{x} + Ax = f(t)$$

with  $(A, B \in \mathbb{C}^{m \times n})$ ,  $f : I \rightarrow \mathbb{C}^m$  a continuous and twice-differentiable function,  $I \subseteq \mathbb{R}$  a compact interval and  $m \in \mathbb{N}$  is called a linear differential-algebraic equation with constant coefficients.

It may be paired with an initial condition

$$x(t_0) = x_0.$$

**Remark.** The equations  $B\dot{x} + Ax = f(t)$  and  $\tilde{B}\dot{\tilde{x}} + \tilde{A}\tilde{x} = \tilde{f}(t)$ , with

$$\begin{aligned} \tilde{A} &= PAQ, & \tilde{B} &= PBQ, & \tilde{f} &= Pf, & \tilde{x} &= Q^{-1}x, \\ P &\in \mathbb{C}^{m \times m}, & Q &\in \mathbb{C}^{n \times n} \end{aligned}$$

and  $P, Q$  nonsingular matrices, are both linear differential-algebraic equations with constant coefficients and the relation  $x = Q\tilde{x}$  provides a one-to-one mapping between their solution sets.

Next, we shall take as examples pencils of matrices made up of the blocks described before in order to show how the linear DAE with constant coefficients associated may be solved.

**Example 3.2.** Consider the pencil

$$\Gamma(\lambda) = \begin{bmatrix} \lambda & 1 & 0 \\ 0 & \lambda & 1 \end{bmatrix}.$$

We write the linear DAE with constant coefficients associated with  $\Gamma(\lambda)$  as

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \\ \dot{x}_3 \end{bmatrix} + \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} f_1(t) \\ f_2(t) \end{bmatrix}.$$

To compute its solutions we must solve the linear system associated with the DAE specified above

$$\begin{cases} \dot{x}_1 + x_2 - f_1(t) = 0 \\ \dot{x}_2 + x_3 - f_2(t) = 0 \end{cases} \quad (1)$$

Denoting with  $g$  a generic  $C^2$  function, the solution is

$$\begin{cases} x_1 = g \\ x_2 = -\dot{x}_1 + f_1(t) \\ x_3 = \ddot{x}_1 - \dot{f}_1(t) + f_2(t) \end{cases}$$



**Example 3.3.** For this example, we'll take the transpose of the pencil of matrices given in example [3.2](#)

$$\Gamma(\lambda) = \begin{bmatrix} \lambda & 0 \\ 1 & \lambda \\ 0 & 1 \end{bmatrix}.$$

The DAE associated with  $\Gamma(\lambda)$  is

$$\begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} + \begin{bmatrix} 0 & 0 \\ 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} f_1(t) \\ f_2(t) \\ f_3(t) \end{bmatrix}.$$

We write this as a linear system

$$\begin{cases} \dot{x}_1 - f_1(t) = 0 \\ \dot{x}_2 + x_1 - f_2(t) = 0 \\ x_2 - f_3(t) = 0 \end{cases}$$

The system can be solved if and only if

$$f_1(t) = \dot{f}_2(t) - \ddot{f}_3(t),$$

and its solution is

$$\begin{cases} x_1 = f_2(t) - \dot{f}_3(t) \\ x_2 = f_3(t) \end{cases}$$

**Example 3.4.** Let  $\Gamma(\lambda)$  be the pencil of matrices

$$\Gamma(\lambda) = \begin{bmatrix} 1 & \lambda & 0 & 0 \\ 0 & 1 & \lambda & 0 \\ 0 & 0 & 1 & \lambda \\ 0 & 0 & 0 & 1 \end{bmatrix}.$$

Rewriting this as a linear DAE with constant coefficients yields

$$\begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \\ \dot{x}_3 \\ \dot{x}_4 \end{bmatrix} + \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} f_1(t) \\ f_2(t) \\ f_3(t) \\ f_4(t) \end{bmatrix}$$

We follow the same steps as in other examples

$$\begin{cases} \dot{x}_2 + x_1 - f_1(t) = 0 \\ \dot{x}_3 + x_2 - f_2(t) = 0 \\ \dot{x}_4 + x_3 - f_3(t) = 0 \\ x_4 - f_4(t) = 0 \end{cases}$$

The solution is

$$\begin{cases} x_1 = f_1(t) - \dot{f}_2(t) + \ddot{f}_3(t) - \dddot{f}_4(t) \\ x_2 = f_2(t) - \dot{f}_3(t) + \ddot{f}_4(t) \\ x_3 = f_3(t) - \dot{f}_4(t) \\ x_4 = f_4(t) \end{cases}$$

**Example 3.5.** Let us take the pencil

$$\Gamma(\lambda) = \begin{bmatrix} \lambda & 1 & 0 & 0 \\ 0 & \lambda & 0 & 0 \\ 0 & 0 & \lambda + 2 & 1 \\ 0 & 0 & 0 & \lambda + 2 \end{bmatrix}$$

and write the DAE associated with it

$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \\ \dot{x}_3 \\ \dot{x}_4 \end{bmatrix} + \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 2 & 1 \\ 0 & 0 & 0 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} f_1(t) \\ f_2(t) \\ f_3(t) \\ f_4(t) \end{bmatrix}.$$

The linear system to solve is

$$\begin{cases} \dot{x}_1 + x_2 - f_1(t) = 0 \\ \dot{x}_2 - f_2(t) = 0 \\ \dot{x}_3 + 2x_3 + x_4 - f_3(t) = 0 \\ \dot{x}_4 + 2x_4 - f_4(t) = 0 \end{cases}$$

Denoting with  $g$  a  $C^2$  function such that

$$\ddot{g} = \dot{f}_1(t) - f_2(t),$$

and (with)  $D, K$  arbitrary constants, the solution is

$$\begin{cases} x_1 = g \\ x_2 = f_1(t) - \dot{g} \\ x_3 = -\frac{1}{4}(4Kt - 2f_3(t)e^{2t} + f_4(t)e^{2t} - 4K)e^{-2t} \\ x_4 = \frac{1}{2}(f_4(t)e^{2t} + 2D)e^{-2t} \end{cases}$$

# Computation of Kronecker's Canonical Form

The following chapter deals with the problem of computing Kronecker's canonical form for a matrix pair  $(A, B)$ . The approach described will be shown to be correct, and an implementation of it using the CAS SageMath has been made publicly available under MIT License on GitHub [\[5\]](#).

We shall divide it into two subproblems: the first one deals with regular pencils of matrices, the other with singular pencils.

At the end of each of the following sections, the steps described shall be summarised in pseudocode.

## Regular pencils.

Let  $\Gamma(\lambda) = A + \lambda B$  be a regular pencil of matrices defined in a vector space over a field  $F$ .

First, let us clarify what we are about to prove.

**Theorem 4.1** (KCF of a regular pencil of matrices). Every regular pencil of matrices  $\Gamma(\lambda) = A + \lambda B$  can be reduced to a matrix of the form

$$\begin{bmatrix} N^{(u_1)} & & & & \\ & N^{(u_2)} & & & \\ & & \ddots & & \\ & & & N^{(u_s)} & \\ & & & & G + \lambda I \end{bmatrix},$$

where the first  $s$  diagonal blocks correspond to infinite elementary divisors  $\mu^{u_1}, \dots, \mu^{u_s}$  of  $\Gamma(\lambda)$  and the last block is uniquely determined by the finite elementary divisors of the given pencil.

*Proof.* The following proof will describe a procedure to reduce  $\Gamma(\lambda)$  to Kronecker's canonical form.

First, we shall find a value  $c \in F$  such that  $\det(\Gamma(c)) \neq 0$  and define the matrix

$$A_1 = A + cB.$$

Now, we rewrite  $\Gamma(\lambda)$  in terms of  $A_1$ ,

$$\Gamma(\lambda) = A_1 + (\lambda - c)B,$$

and premultiply it by  $A_1^{-1}$

$$A_1^{-1}\Gamma(\lambda) = I + (\lambda - c)A_1^{-1}B. \quad (2)$$

Let us denote with  $J$  the JCF of  $A_1^{-1}B$  and  $P_1$  the similarity matrix used to compute it

$$J = P_1^{-1}(A_1^{-1}B)P_1,$$

and let us assume it is of the form

$$J = \begin{bmatrix} J_1 & \\ & J_0 \end{bmatrix}$$

with  $J_0$  a nilpotent Jordan matrix and  $J_1$  such that  $\det(J_1) \neq 0$  (which is the form returned by SageMath when computing Jordan matrices); call  $j, k$  the number of rows in  $J_0$  and  $J_1$  respectively.

Define the permutation matrix

$$P_\pi = \begin{bmatrix} & I^{(k)} \\ I^{(j)} & \end{bmatrix}$$

so that we can write

$$P_\pi^T J P_\pi = J' = \begin{bmatrix} J_0 & \\ & J_1 \end{bmatrix}.$$

Now, we rewrite (2) substituting  $A_1^{-1}B$  with its permuted Jordan form

$$A_1^{-1}\Gamma(\lambda) = I + (\lambda - c)P_1 P_\pi^{-T} J' P_\pi^{-1} P_1^{-1}. \quad (3)$$

We can write the identity matrix in terms of  $P_1$  as  $I = P_1 P_1^{-1}$  in (3)

$$\begin{aligned} A_1^{-1}\Gamma(\lambda) &= \\ &= P_1 P_1^{-1} + (\lambda - c)P_1 P_\pi^{-T} J' P_\pi^{-1} P_1^{-1} = \\ &= P_1 (P_1^{-1} + (\lambda - c)P_\pi^{-T} J' P_\pi^{-1} P_1^{-1}). \end{aligned}$$

We premultiply it by  $P_1^{-1}$

$$P_1^{-1} A_1^{-1} \Gamma(\lambda) = P_1^{-1} + (\lambda - c)P_\pi^{-T} J' P_\pi^{-1} P_1^{-1}$$

and then postmultiply by  $P_1$

$$P_1^{-1} A_1^{-1} \Gamma(\lambda) P_1 = I + (\lambda - c) P_\pi^{-T} J' P_\pi^{-1}.$$

The very same steps can be followed for  $P_\pi^{-T}$  and  $P_\pi^{-1}$ .  
The final result is

$$P_\pi^T P_1^{-1} A_1^{-1} \Gamma(\lambda) P_1 P_\pi = I + (\lambda - c) J'. \quad (4)$$

Now, we may work on the blocks  $J_0, J_1$  of  $J'$ .

For the following steps, it is useful to rewrite the expression on the right side of (4) so that the form of its diagonal blocks is explicitly readable, meaning

$$I + (\lambda - c) J' = \begin{bmatrix} I^{(j)} - cJ_0 + \lambda J_0 & \\ & I^{(k)} - cJ_1 + \lambda J_1 \end{bmatrix}.$$

Let us start from the first diagonal block

$$I^{(j)} - cJ_0 + \lambda J_0.$$

First, we need to postmultiply it in (4) by  $K = (I^{(j)} - cJ_0)^{-1}$

$$P_\pi^T P_1^{-1} A_1^{-1} \Gamma(\lambda) P_1 P_\pi \begin{bmatrix} K & \\ & I^{(k)} \end{bmatrix} = \begin{bmatrix} I^{(j)} + \lambda K J_0 & \\ & I^{(k)} - cJ_1 + \lambda J_1 \end{bmatrix}.$$

Let us denote with  $H^{(j)}$  the JCF of  $K J_0$ ,  $P_2$  the similarity matrix used to compute it and  $N^{(j)} = I^{(j)} + \lambda H^{(j)}$

$$N^{(j)} = I^{(j)} + \lambda H^{(j)} = I^{(j)} + \lambda P_2^{-1} (K J_0) P_2.$$

**Remark.**  $H$  is an upper shift matrix as it is the JCF of a nilpotent matrix.

Following the analogous steps to handle  $P_2$  and its inverse yields us

$$\begin{bmatrix} P_2^{-1} & \\ & I \end{bmatrix} P_\pi^T P_1^{-1} A_1^{-1} \Gamma(\lambda) P_1 P_\pi \begin{bmatrix} K P_2 & \\ & I^{(k)} \end{bmatrix} = \begin{bmatrix} N^{(j)} & \\ & I^{(k)} - cJ_1 + \lambda J_1 \end{bmatrix}. \quad (5)$$

At this point, we can focus on the second diagonal block

$$I^{(k)} - cJ_1 + \lambda J_1.$$

We postmultiply it in (5) by  $J_1^{-1}$

$$\begin{bmatrix} P_2^{-1} & \\ & I \end{bmatrix} P_\pi^T P_1^{-1} A_1^{-1} \Gamma(\lambda) P_1 P_\pi \begin{bmatrix} K P_2 & \\ & J_1^{-1} \end{bmatrix} = \begin{bmatrix} N^{(j)} & \\ & J_1^{-1} + (\lambda - c)I^{(k)} \end{bmatrix}. \quad (6)$$

Let us denote with  $G$  the JCF of the constant term  $J_1^{-1} - cI^{(k)}$  and  $P_3$  the similarity matrix used to compute it

$$J_1^{-1} - cI^{(k)} = P_3^{-1} G P_3.$$

Again, we follow the same procedure to handle  $P_3$  and its inverse, thus obtaining our end result

$$\begin{bmatrix} P_2^{-1} & \\ & P_3^{-1} \end{bmatrix} P_\pi^T P_1^{-1} A_1^{-1} \Gamma(\lambda) P_1 P_\pi \begin{bmatrix} K P_2 & \\ & J_1^{-1} P_3 \end{bmatrix} = \begin{bmatrix} N^{(j)} & \\ & G + \lambda I^{(k)} \end{bmatrix}. \quad (7)$$

□

To conclude, we shall present the aforementioned steps used in order to compute Kronecker's canonical form  $K(\Gamma(\lambda))$  in pseudocode.

---

**Algorithm 1:** Procedure to compute KCF of a regular pencil.

---

**Data:**  $\Gamma(\lambda) = A + \lambda B$ : regular pencil

**Result:**  $K$ : KCF of the pencil of matrices  $\Gamma(\lambda)$

**while** *True* **do**

$c \leftarrow$  random value;

**if**  $\det(\Gamma(c)) \neq 0$  **then**

**break**

**end**

**end**

$A_1 \leftarrow A + c * B$ ;

$J \leftarrow \text{jordan}(A_1^{-1} * B)$ ;

$\{J_0, J_1\} \leftarrow \text{submatrices}(J)$ ;

$\{N_i\}_{i \geq 0} \leftarrow \text{jordan}((I - c * J_0)^{-1} * J_0)$ ;

$G \leftarrow \text{jordan}(J_1^{-1} - c * I)$ ;

$K \leftarrow \text{diag}(\{N_i\}, G + \lambda I)$ ;

**return**  $K$ ;

---

## Singular pencils.

Let  $\Gamma(\lambda)$  be a singular pencil of  $m \times n$  matrices the entries of which lie in a field  $F$

$$\Gamma(\lambda) = A + \lambda B.$$

The pencil being singular implies either  $r < m$  or  $r < n$  with  $r$  rank of the pencil in the field  $F(\lambda)$  of rational functions; let us assume without loss of generality the relation  $r < n$  holds: otherwise, we shall work with the transpose of the pencil of matrices  $\Gamma^T(\lambda) = A^T + \lambda B^T$  in order to reduce the problem to the case we shall describe hereafter.

First, we shall find a polynomial  $x(\lambda)$  of minimal degree  $\epsilon$  in the (right) kernel of  $\Gamma(\lambda)$

$$x(\lambda) = x_0 - \lambda x_1 + \lambda^2 x_2 + \dots + (-1)^\epsilon \lambda^\epsilon x_\epsilon.$$

In order to define a procedure to compute it, we consider the family of matrices  $M_k^{(A,B)}$  defined as

$$M_0^{(A,B)} = \begin{bmatrix} A \\ B \end{bmatrix}, \quad M_1^{(A,B)} = \begin{bmatrix} A & 0 \\ B & A \\ 0 & B \end{bmatrix}, \quad M_k^{(A,B)} = \begin{bmatrix} A & 0 & \dots & 0 \\ B & A & & \vdots \\ 0 & B & \ddots & \\ \vdots & \vdots & \ddots & A \\ 0 & 0 & \dots & B \end{bmatrix}.$$

The basis matrix of the (right) kernel of a matrix  $M_k$  - split into blocks of size  $k + 1$  - identifies the polynomial  $g(\lambda)$  of degree  $k$  which satisfies the equations

$$Ag_0 = 0, \quad Bg_0 - Ag_1 = 0, \quad \dots, \quad Bg_{k-1} - Ag_k = 0, \quad Bg_k = 0.$$

Note that if there is no polynomial of degree  $k$  in the kernel of the pencil  $\Gamma(\lambda)$ , then  $\ker(M_k)$  is empty.

Iteratively building the succession of matrices  $\{M_i\}_{i \geq 0}$  and computing their kernels yields us the polynomial  $x(\lambda)$ , which is what we were looking for.

Depending on the value of  $\epsilon$ , we shall distinguish two cases.



**Case 1.**  $\epsilon > 0$ .

We want to prove the following theorem.

**Theorem 4.2** (Reduction theorem). Given a pencil of matrices  $\Gamma(\lambda)$ , if the polynomial of minimal degree  $\epsilon$  in the (right) kernel of  $\Gamma(\lambda)$  has degree  $\epsilon > 0$ , then  $\Gamma(\lambda)$  is equivalent to a pencil of the form

$$\begin{bmatrix} L_\epsilon & 0 \\ 0 & A^* + \lambda B^* \end{bmatrix}.$$

*Proof.* We shall prove the theorem by defining a procedure to compute such a transformation.

We know both the vectors in  $B_1, B_2$  are linearly independent (a reader shall refer to Gantmacher for the proof), with

$$B_1 = \{x_0, x_1, \dots, x_\epsilon\}, \quad B_2 = \{Ax_0, Ax_1, \dots, Ax_\epsilon\}.$$

Denote with  $P, Q$  the matrices obtained by completing the vectors in  $B_2, B_1$  to a basis in  $F^m, F^n$  respectively; applying a change of basis to the pencil of matrices  $\Gamma(\lambda)$  gives us

$$\tilde{\Gamma}(\lambda) = \tilde{A} + \lambda \tilde{B} = P^{-1}AQ + \lambda P^{-1}BQ,$$

with  $\tilde{A}, \tilde{B}$  of the kind

$$\tilde{A} = \begin{bmatrix} \boxed{H^{(\epsilon)} \quad O} & * \dots * \end{bmatrix}, \quad \tilde{B} = \begin{bmatrix} \boxed{I^{(\epsilon)} \quad O} & * \dots * \end{bmatrix};$$

the notation  $* \dots *$  is used for a block with no relevant structure,  $O$  is a null (column) vector of length  $\epsilon$ .

For the sake of clarity, we shall partition the pencil of matrices  $\tilde{\Gamma}(\lambda)$  as

$$\begin{bmatrix} L_\epsilon & D + \lambda F \\ 0 & A^* + \lambda B^* \end{bmatrix}$$

with  $L_\epsilon$  the first rectangular  $\epsilon \times (\epsilon + 1)$  block.

Now we need to transform the pencil of matrices  $\tilde{\Gamma}(\lambda)$  so that it is upper triangular.

We shall do this via the transformation

$$\begin{bmatrix} I^{(\epsilon)} & Y \\ 0 & I^{(m-\epsilon)} \end{bmatrix} \begin{bmatrix} L_\epsilon & D + \lambda F \\ 0 & A^* + \lambda B^* \end{bmatrix} \begin{bmatrix} I^{(\epsilon+1)} & -X \\ 0 & I^{(n-\epsilon-1)} \end{bmatrix}. \quad (8)$$

Applying the transformation in (8) yields

$$\begin{bmatrix} L_\epsilon & D + \lambda F + Y(A^* + \lambda B^*) - L_\epsilon X \\ 0 & A^* + \lambda B^* \end{bmatrix}.$$

In other words, to make the pencil  $\tilde{\Gamma}(\lambda)$  upper triangular, we shall determine two matrices  $X, Y$  such that

$$D + \lambda F + Y(A^* + \lambda B^*) - L_\epsilon X = 0.$$

Denoting with  $\mathbf{y}_i, \mathbf{a}_i, \mathbf{b}_i$  the  $i$ -th row of  $Y$ , column of  $A$  and  $B$  respectively, we shall rewrite this condition as a system of linear equations

$$\begin{cases} x_{2,k} + \lambda x_{1,k} = d_{1,k} + \lambda f_{1,k} + \mathbf{y}_1 \mathbf{a}_k + \lambda \mathbf{y}_1 \mathbf{b}_k \\ x_{3,k} + \lambda x_{2,k} = d_{2,k} + \lambda f_{2,k} + \mathbf{y}_2 \mathbf{a}_k + \lambda \mathbf{y}_2 \mathbf{b}_k \\ \dots \\ x_{\epsilon+1,k} + \lambda x_{\epsilon,k} = d_{\epsilon,k} + \lambda f_{\epsilon,k} + \mathbf{y}_\epsilon \mathbf{a}_k + \lambda \mathbf{y}_\epsilon \mathbf{b}_k \end{cases} \quad (9)$$

$$(k = 1, \dots, n - \epsilon - 1)$$

To approach this problem, we shall define the  $1 \times (n - \epsilon - 1)$  matrix  $W$  as the differences between subsequent rows in  $D$  and  $F$  flipping the sign every  $m - \epsilon - 1$  elements, but first let us differentiate between two other subcases.

To make this definition easier to understand, we shall present the pseudocode of an algorithm to compute  $W$ . Now, we shall distinguish two other subcases.

*Subcase (i):  $\epsilon = 1$ .*

In this case, there are no conditions on the values of the elements of  $Y$ ; we shall choose arbitrary values for its entries and compute those in  $X$  according to the last equation in the linear system in (9).

*Subcase (ii):  $\epsilon > 1$ .*

In this case, we shall define the matrix  $W$  according to the definition given above; in an attempt to make the explanation easier to understand, we shall give the pseudocode.

---

**Algorithm 2:** Procedure to compute  $W$ .

---

```

sign  $\leftarrow -1$ ;
 $W \leftarrow []$ ;
i  $\leftarrow 0$ ;
for  $i < n - \epsilon - 1$  do
    j  $\leftarrow 0$ ;
    for  $j < m - \epsilon - 1$  do
        if  $j \bmod (m - \epsilon - 1) = 0$  then
            sign  $\leftarrow sign * (-1)$ ;
        end
        append( $W$ ,  $sign * (f_{i+1,j} - d_{i,j})$ );
    end
end
return matrix( $W$ )

```

---

Solving for  $Z$  in  $ZM_{\epsilon-2}^{(A^*, B^*)} = W$  yields a matrix of the form

$$Z^T = \begin{bmatrix} \mathbf{y}_1 - \mathbf{y}_2 \\ -(\mathbf{y}_2 - \mathbf{y}_3) \\ \dots \\ \mathbf{y}_{\epsilon-1} - \mathbf{y}_\epsilon \end{bmatrix}.$$

We can now recover  $Y$  and, subsequently,  $X$  from the linear system in (9).  $\square$

To help make the procedure more intelligible, we shall present the pseudocode to apply the transformation in [Reduction theorem](#) (theorem 4.2).

---

**Algorithm 3:** Procedure to compute KCF of a singular pencil.

---

**Data:**  $\Gamma(\lambda) = A + \lambda B$ : singular pencil  
**Result:**  $K$ : KCF of the pencil of matrices  $\Gamma(\lambda)$   
 $V \leftarrow$  polynomial of minimal degree  $\in \ker(\Gamma)$ ;  
 $Q \leftarrow$  complete to a basis( $V$ );  
 $P \leftarrow$  complete to a basis( $AV$ );  
 $\tilde{A} = P^{-1}AQ$ ;  
 $\tilde{B} = P^{-1}BQ$ ;  
 $L_\epsilon, D, F, A^*, B^* \leftarrow$  partition pencil( $\tilde{A} + \lambda\tilde{B}$ );  
**if**  $\text{degree}(V) \neq 0$  **then**  
     $M \leftarrow$  build  $M_\epsilon(A^*, B^*, \epsilon - 1)$ ;  
     $W \leftarrow$  build  $W(F, D)$ ;  
     $Z \leftarrow WM^{-1}$ ;  
     $Y \leftarrow$  partition( $Z$ );  
     $X \leftarrow$  solve system in (9);  
     $L \leftarrow$  block matrix( $[[I, Y], [0, I]]$ );  
     $R \leftarrow$  block matrix( $[[I, -X], [0, I]]$ );  
    **return**  $LP^{-1}\Gamma(\lambda)QR$ ;  
**else**  
    **return**  $P^{-1}\Gamma(\lambda)Q$ ;  
**end**

---

**Case 2.**  $\epsilon = 0$ .

In this case, the columns of the pencil are connected by linear relations with constant coefficients.

This implies the transformation given by matrices  $P, Q$  returns a pencil of the form

$$\begin{bmatrix} 0 & A^* + \lambda B^* \end{bmatrix},$$

and this procedure can be conducted as many times as the number of independent constant solutions of the equation  $(A + \lambda B)x = 0$  so that we may reduce the problem to **Case 1**.

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