# The Chevalley-Jordan Decomposition and Spectral Projections of Complex Matrices 

Said Zriaa and Mohammed Mouçouf


#### Abstract

In this paper, a novel and simple method for obtaining the Chevalley-Jordan decomposition and the spectral projections of matrices is presented. Our method is direct and elementary, it gives tractable and manageable formulas with minimum mathematical prerequisites. Moreover, knowing only some associated matrices of the matrix, we can simply provide the minimal polynomial of this matrix.


Key Words: The Chevalley-Jordan decomposition, spectral projections, minimal polynomial.

## Contents

1 Introduction

2 Main results 2
3 Conclusion 5

## 1. Introduction

The Chevalley-Jordan decomposition of a complex matrix plays a central role in linear algebra. In a recent work, the Chevalley-Jordan decomposition is proved to be a powerful technique in electroencephalography signals [1]. This decomposition is a further development of Jordan canonical form. For any square complex matrix the Chevalley-Jordan decomposition exists and unique, it expresses any square complex matrix as the sum of its commuting diagonalizable part and nilpotent part. This decomposition can be described if the matrix has its Jordan normal form (which is practically difficult to compute), but it may computed even if the Jordan normal form does not.

The main purpose of this work is to propose a new and simple method for determining explicitly the Chevalley-Jordan decomposition and the spectral projections of matrices in a direct way.

Let $\alpha_{1}, \alpha_{2}, \ldots, \alpha_{s}$ be distinct elements of $\mathbb{C}$ and $m_{1}, m_{2}, \ldots, m_{s}$ be nonnegative integers. For any polynomial $P(x)=\left(x-\alpha_{1}\right)^{m_{1}}\left(x-\alpha_{2}\right)^{m_{2}} \cdots\left(x-\alpha_{s}\right)^{m_{s}}$, we denote by $L_{j k_{j}}(x)[P]$ the following polynomial

$$
\begin{equation*}
L_{j k_{j}}(x)[P]=P_{j}(x)\left(x-\alpha_{j}\right)^{k_{j}} \sum_{i=0}^{m_{j}-1-k_{j}} \frac{1}{i!} g_{j}^{(i)}\left(\alpha_{j}\right)\left(x-\alpha_{j}\right)^{i} \tag{1.1}
\end{equation*}
$$

where $1 \leq j \leq s, 0 \leq k_{j} \leq m_{j}-1$,

$$
P_{j}(x)=\prod_{i=1, i \neq j}^{s}\left(x-\alpha_{i}\right)^{m_{i}}=\frac{P(x)}{\left(x-\alpha_{j}\right)^{m_{j}}}, 1 \leq j \leq s
$$

and

$$
g_{j}(x)=\left(P_{j}(x)\right)^{-1}
$$

Here $g_{j}^{(i)}(x)$ denotes the $i$ th derivative of $g_{j}(x)$.
These polynomials are of great importance, they are used in [4] to invert the confluent Vandermonde matrix and used in [5] for computing the exponential of complex matrices.

[^0]
## 2. Main results

In this section, we will be interested in determining explicit formulas for the Chevalley-Jordan decomposition and the spectral projections of complex matrices. Our contribution does not consist only in giving these formulas, but also in making their determination much practical and expressing them in an elegant representations.
Now, we start by formulating the following important result.
Theorem 2.1. Let $A$ be a $k \times k$ matrix, and let $\chi_{A}(x)=\left(x-\alpha_{1}\right)^{m_{1}}\left(x-\alpha_{2}\right)^{m_{2}} \cdots\left(x-\alpha_{s}\right)^{m_{s}}, \alpha_{1}=0$, be its characteristic polynomial (possibly $m_{1}=0$ ). Then for every $n \in \mathbb{N}$, we have

$$
\begin{equation*}
A^{n}=\sum_{j=0}^{m_{1}-1} \delta_{n j} B_{1 j}+\sum_{j=0}^{m_{2}-1}\binom{n}{j} \alpha_{2}^{n-j} B_{2 j}+\cdots+\sum_{j=0}^{m_{s}-1}\binom{n}{j} \alpha_{s}^{n-j} B_{s j} \tag{2.1}
\end{equation*}
$$

where $B_{j k_{j}}=L_{j k_{j}}(A)\left[\chi_{A}\right]$ and $\delta_{n j}$ denotes the Kronecker symbol.
Proof. Using Theorem 2.9 of [5], one can easily prove this result.
In the present approach, one gains a basic construction of the so-called spectral decomposition of A. The approach makes the determination of the spectral decomposition of any square matrix more practical than the usual method of partial fraction decomposition. Using Lagrange polynomials W. A. Harris et al. [2] have derived the spectral decomposition of a matrix with simple eigenvalues. Here, we generalize this result to any matrix using a generalization of Hermite's interpolation formula given by A. Spitzbart [6].

Using the previous theorem we find a new method to calculate the Chevalley-Jordan decomposition and the spectral projections of $A$ at the same time.

Theorem 2.2. Let $A$ be a $k \times k$ matrix, and let $\chi_{A}(x)=\left(x-\alpha_{1}\right)^{m_{1}}\left(x-\alpha_{2}\right)^{m_{2}} \cdots\left(x-\alpha_{s}\right)^{m_{s}}, \alpha_{1}=0$, be its characteristic polynomial (possibly $m_{1}=0$ ). Then the Chevalley-Jordan decomposition of $A$ is

$$
\begin{equation*}
A=D+N \tag{2.2}
\end{equation*}
$$

where

$$
\begin{equation*}
N=B_{11}+B_{21}+\cdots+B_{s 1} \tag{2.3}
\end{equation*}
$$

and

$$
\begin{equation*}
\boldsymbol{D}=\alpha_{2} B_{20}+\alpha_{3} B_{30}+\cdots+\alpha_{s} B_{s 0} \tag{2.4}
\end{equation*}
$$

where $B_{j r}=L_{j r}(A)\left[\chi_{A}\right]$ for $j=1,2, \ldots, s$ and $r=0,1$. Moreover, $B_{10}, B_{20}, \ldots, B_{s 0}$ are the spectral projections of $A$ at $\alpha_{1}, \alpha_{2}, \ldots, \alpha_{s}$, respectively.

Proof. To obtain

$$
A=D+N
$$

and

$$
\begin{equation*}
I=B_{10}+B_{20}+\cdots+B_{s 0} \tag{2.5}
\end{equation*}
$$

it suffices to take $n=1$ and $n=0$ in Formula (2.1), respectively. Since $B_{j 0}, B_{j 1} ; j=1,2, \ldots, s$, are polynomials of the matrix $A$, we have $D N=N D$.

On the other hand, it is clear that $B_{j 0} B_{i 0}=B_{i 0} B_{j 0}=0, i \neq j$. Multiplying both sides of (2.5) by $B_{j 0}$, we get $B_{j 0}^{2}=B_{j 0}, j=1,2, \ldots, s$.

Furthermore, the matrix $B_{j 0}$ is diagonalizable, then so is $\alpha_{j} B_{j 0}$. The fact that $\alpha_{j} B_{j 0}$ and $\alpha_{i} B_{i 0}$ commute assures then that $D$ is diagonalizable.

To complete the proof, it remains to show that the matrix $N$ is nilpotent. To see this, it suffices to verify, using Formula (1.1), that each $B_{j 1}$ is nilpotent.

Remark 2.3. We can immediately find that $B_{j 0}, j=1,2, \ldots, s$, are the spectral projections of $A$ by using Proposition 3.1 and Corollary 3.8 of [3].

The following examples are illustrations of Theorem 2.2.
Example 2.4. Let us consider the following matrix

$$
A=\left(\begin{array}{lll}
2 & 0 & 1 \\
0 & 2 & 0 \\
0 & 0 & 3
\end{array}\right)
$$

It is clear that the characteristic polynomial of $A$ is $\chi_{A}(x)=(x-2)^{2}(x-3)$. The Chevalley-Jordan decomposition of this matrix is

$$
A=D+N
$$

where

$$
D=2 B_{10}+3 B_{20} \quad \text { and } \quad N=B_{11}
$$

A trivial calculation yields

$$
\begin{aligned}
& B_{10}=L_{10}(A)\left[\chi_{A}\right]=\left(\begin{array}{ccc}
1 & 0 & -1 \\
0 & 1 & 0 \\
0 & 0 & 0
\end{array}\right) \\
& B_{11}=L_{11}(A)\left[\chi_{A}\right]=\left(\begin{array}{lll}
0 & 0 & 0 \\
0 & 0 & 0 \\
0 & 0 & 0
\end{array}\right) \\
& B_{20}=L_{20}(A)\left[\chi_{A}\right]=\left(\begin{array}{lll}
0 & 0 & 1 \\
0 & 0 & 0 \\
0 & 0 & 1
\end{array}\right)
\end{aligned}
$$

Then

$$
D=\left(\begin{array}{lll}
2 & 0 & 1 \\
0 & 2 & 0 \\
0 & 0 & 3
\end{array}\right)=A \text { and } N=0
$$

We deduce that $A$ is diagonalizable.
Example 2.5. Consider the matrix

$$
A=\left(\begin{array}{cccc}
1 & -1 & 1 & -1 \\
0 & -2 & 0 & -3 \\
2 & 3 & 0 & 3 \\
1 & 5 & -1 & 6
\end{array}\right)
$$

The characteristic polynomial of $A$ is $\chi_{A}(x)=(x-2)(x-1)^{3}$. The Chevalley-Jordan decomposition of this matrix is

$$
A=D+N
$$

where

$$
\boldsymbol{D}=2 B_{10}+B_{20} \quad \text { and } \quad \boldsymbol{N}=B_{21}
$$

On the other hand, we have

$$
\left\{\begin{array}{l}
B_{10}=(A-I)^{3} \\
B_{20}=(A-2 I)\left(-A^{2}+A-I\right) \\
B_{21}=-A(A-2 I)(A-I)
\end{array}\right.
$$

Simple calculation gives

$$
\begin{aligned}
B_{10} & =\left(\begin{array}{llll}
1 & 1 & 0 & 1 \\
0 & 0 & 0 & 0 \\
1 & 1 & 0 & 1 \\
0 & 0 & 0 & 0
\end{array}\right) \\
B_{20} & =\left(\begin{array}{cccc}
0 & -1 & 0 & -1 \\
0 & 1 & 0 & 0 \\
-1 & -1 & -1 & -1 \\
0 & 0 & 0 & 1
\end{array}\right) \\
B_{21} & =\left(\begin{array}{cccc}
-1 & -2 & 1 & -2 \\
0 & -3 & 0 & -3 \\
1 & 2 & -1 & 2 \\
1 & 5 & -1 & 5
\end{array}\right)
\end{aligned}
$$

Then

$$
\boldsymbol{D}=\left(\begin{array}{cccc}
2 & 1 & 0 & 1 \\
0 & 1 & 0 & 0 \\
1 & 1 & -1 & 1 \\
0 & 0 & 0 & 1
\end{array}\right)
$$

and

$$
\boldsymbol{N}=\left(\begin{array}{cccc}
-1 & -2 & 1 & -2 \\
0 & -3 & 0 & -3 \\
1 & 2 & -1 & 2 \\
1 & 5 & -1 & 5
\end{array}\right)
$$

The following Theorem shows that knowing only the associated matrices $B_{i j}$ of $A$, we can simply provide the minimal polynomial of the matrix $A$.

Theorem 2.6. Let $A$ be a matrix and let $\chi_{A}(x)=\left(x-\alpha_{1}\right)^{m_{1}}\left(x-\alpha_{2}\right)^{m_{2}} \cdots\left(x-\alpha_{s}\right)^{m_{s}}$ be its characteristic polynomial with $\alpha_{1}=0$ (possibly $m_{1}=0$ ). Then,

1. The index of $\alpha_{i}$ is the greatest integer $j$ such that $B_{i j-1} \neq 0$.
2. The index of $\alpha_{i}$ is 1 if and only if $B_{i 1}=0$.
3. $A$ is diagonalizable if and only if $B_{i 1}=0, i=1,2, \ldots, s$.

Proof. Clearly Formula (2.1) is a $\mathcal{P}$-canonical form of $A$ (see [3]) and the result follows then from Corollary 3.6., Theorem 4.1. of [3] and Theorem 2.1 above.

Let us consider the same matrix of Example (2.5)
Example 2.7. We have

$$
A=\left(\begin{array}{cccc}
1 & -1 & 1 & -1 \\
0 & -2 & 0 & -3 \\
2 & 3 & 0 & 3 \\
1 & 5 & -1 & 6
\end{array}\right)
$$

Since $B_{21} \neq 0$, then using (3) of the last theorem the matrix $A$ is not diagonalizable.
It is easy to verify that

$$
B_{22}=-(A-2 I)(A-I)^{2}=\left(\begin{array}{cccc}
0 & 0 & 0 & 0 \\
-3 & -6 & 3 & -6 \\
0 & 0 & 0 & 0 \\
3 & 6 & -3 & 6
\end{array}\right)
$$

using (1) of the last theorem the minimal polynomial of $A$ is $(x-2)(x-1)^{3}$.

## 3. Conclusion

We have presented a new and elegant method to facilitate the computation of the Chevalley-Jordan decomposition and the spectral projections of matrices. We can also provide the minimal polynomial of matrices.

## Acknowledgments

The authors would like to thank the referees for their helpful comments on the original version of this paper.

## References

1. A. A. Ahmad Fuad and T. Ahmad, Decomposing the Krohn-Rhodes form of electroencephalography (EEG) signals using Jordan-Chevalley decomposition technique, Axioms, vol. 10, no. 1, p. 10, (2021).
2. W. A. Harris, J. P. Fillmore, and D. R. Smith, Matrix Exponentials-Another Approach. SIAM review 43, 694-706 (2001).
3. M. Mouçouf, $\mathcal{P}$-canonical forms and Drazin inverses, arXiv:2007.10199v5 [math.RA](2021).
4. M. Mouçouf, S. Zriaa, A new approach for computing the inverse of confluent Vandermonde matrices via Taylor's expansion, Linear Multilinear Algebra, 70:20, 5973-5986 (2022). DOI:10.1080/03081087.2021.1940807.
5. M. Mouçouf, S. Zriaa, Explicit formulas for the matrix exponential, Boletim da Sociedade Paranaense de Matemática, 41, 1-14 (2023). DOI:10.5269/bspm. 63692.
6. A. Spitzbart, A generalization of Hermite's interpolation formula, Amer. Math. Monthly. 67(1), 42-46, (1960).
[^1]
[^0]:    2010 Mathematics Subject Classification: 15A21, 15A23, 11C99.
    Submitted December 21, 2022. Published March 10, 2023

[^1]:    Said Zriaa and Mohammed Mouçouf,
    Department of Mathematics,
    Faculty of Science, Chouaib Doukkali, Morocco.
    E-mail address: saidzriaa1992@gmail.com
    E-mail address: moucouf@hotmail.com

