

## 10.1 Jordan Form and Eigenanalysis

### Generalized Eigenanalysis

The main result is **Jordan's decomposition**

$$A = PJP^{-1},$$

valid for any real or complex square matrix  $A$ . We describe here how to compute the invertible matrix  $P$  of generalized eigenvectors and the upper triangular matrix  $J$ , called a **Jordan form** of  $A$ .

**Jordan block.** An  $m \times m$  upper triangular matrix  $B(\lambda, m)$  is called a **Jordan block** provided all  $m$  diagonal elements are the same eigenvalue  $\lambda$  and all super-diagonal elements are one:

$$B(\lambda, m) = \begin{pmatrix} \lambda & 1 & 0 & \cdots & 0 & 0 \\ 0 & \lambda & 1 & \cdots & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & \cdots & \lambda & 1 \\ 0 & 0 & 0 & \cdots & 0 & \lambda \end{pmatrix} \quad (m \times m \text{ matrix})$$

**Jordan form.** Given an  $n \times n$  matrix  $A$ , a **Jordan form**  $J$  for  $A$  is a block diagonal matrix

$$J = \mathbf{diag}(B(\lambda_1, m_1), B(\lambda_2, m_2), \dots, B(\lambda_k, m_k)),$$

where  $\lambda_1, \dots, \lambda_k$  are eigenvalues of  $A$  (duplicates possible) and  $m_1 + \dots + m_k = n$ . The eigenvalues of  $J$  are on the diagonal of  $J$  and  $J$  has exactly  $k$  eigenpairs. If  $k < n$ , then  $J$  is non-diagonalizable. Relation  $AP = PJ$  implies  $A$  has exactly  $k$  eigenpairs and  $A$  fails to be diagonalizable for  $k < n$ .

The relation  $A = PJP^{-1}$  is called a **Jordan decomposition** of  $A$ . Invertible matrix  $P$  is called the **matrix of generalized eigenvectors** of  $A$ . It defines a coordinate system  $\vec{x} = P\vec{y}$  in which the vector function  $\vec{x} \rightarrow A\vec{x}$  is transformed to the simpler vector function  $\vec{y} \rightarrow J\vec{y}$ .

If equal eigenvalues are adjacent in  $J$ , then Jordan blocks with equal diagonal entries will be adjacent. Zeros can appear on the super-diagonal of  $J$ , because adjacent Jordan blocks join on the super-diagonal with a zero. A complete specification of how to build  $J$  from  $A$  appears below.

**Decoding a Jordan Decomposition**  $A = PJP^{-1}$ . If  $J$  is a single Jordan block,  $J = B(\lambda, m)$ , then  $P = \langle \vec{v}_1 | \dots | \vec{v}_m \rangle$  and  $AP = PJ$

means

$$\begin{aligned} A\vec{v}_1 &= \lambda\vec{v}_1, \\ A\vec{v}_2 &= \lambda\vec{v}_2 + \vec{v}_1, \\ &\vdots \\ A\vec{v}_m &= \lambda\vec{v}_m + \vec{v}_{m-1}. \end{aligned}$$

The exploded view of the relation  $AP = PB(\lambda, m)$  is called a **Jordan chain**. The formulas can be compacted via matrix  $N = A - \lambda I$  into the recursion

$$N\vec{v}_1 = \vec{0}, \quad N\vec{v}_2 = \vec{v}_1, \dots, N\vec{v}_m = \vec{v}_{m-1}.$$

The first vector  $\vec{v}_1$  is an eigenvector. The remaining vectors  $\vec{v}_2, \dots, \vec{v}_m$  are **not eigenvectors**, they are called **generalized eigenvectors**. A similar formula can be written for each distinct eigenvalue of a matrix  $A$ . The collection of formulas are called **Jordan chain relations**. A given eigenvalue may appear multiple times in the chain relations, due to the appearance of two or more Jordan blocks with the same eigenvalue.

### Theorem 1 (Jordan Decomposition)

Every  $n \times n$  matrix  $A$  has a Jordan decomposition  $A = PJP^{-1}$ .

**Proof:** The result holds by default for  $1 \times 1$  matrices. Assume the result holds for all  $k \times k$  matrices,  $k < n$ . The proof proceeds by induction on  $n$ .

The induction assumes that for any  $k \times k$  matrix  $A$ , there is a Jordan decomposition  $A = PJP^{-1}$ . Then the columns of  $P$  satisfy Jordan chain relations

$$A\vec{x}_i^j = \lambda_i\vec{x}_i^j + \vec{x}_i^{j-1}, \quad j > 1, \quad A\vec{x}_i^1 = \lambda_i\vec{x}_i^1.$$

Conversely, if the Jordan chain relations are satisfied for  $k$  independent vectors  $\{\vec{x}_i^j\}$ , then the vectors form the columns of an invertible matrix  $P$  such that  $A = PJP^{-1}$  with  $J$  in Jordan form. The induction step centers upon producing the chain relations and proving that the  $n$  vectors are independent.

Let  $B$  be  $n \times n$  and  $\lambda_0$  an eigenvalue of  $B$ . The Jordan chain relations hold for  $A = B$  if and only if they hold for  $A = B - \lambda_0 I$ . Without loss of generality, we can assume 0 is an eigenvalue of  $B$ .

Because  $B$  has 0 as an eigenvalue, then  $p = \dim(\mathbf{kernel}(B)) > 0$  and  $k = \dim(\mathbf{Image}(B)) < n$ , with  $p + k = n$ . If  $k = 0$ , then  $B = 0$ , which is a Jordan form, and there is nothing to prove. Assume henceforth  $p$  and  $k$  positive. Let  $S = \langle \mathbf{col}(B, i_1) | \dots | \mathbf{col}(B, i_k) \rangle$  denote the matrix of pivot columns  $i_1, \dots, i_k$  of  $B$ . The pivot columns are known to span  $\mathbf{Image}(B)$ . Let  $A$  be the  $k \times k$  basis representation matrix defined by the equation  $BS = SA$ , or equivalently,  $B \mathbf{col}(S, j) = \sum_{i=1}^k a_{ij} \mathbf{col}(S, i)$ . The induction hypothesis applied to  $A$  implies there is a basis of  $k$ -vectors satisfying Jordan chain relations

$$A\vec{x}_i^j = \lambda_i\vec{x}_i^j + \vec{x}_i^{j-1}, \quad j > 1, \quad A\vec{x}_i^1 = \lambda_i\vec{x}_i^1.$$

The values  $\lambda_i, i = 1, \dots, p$ , are the distinct eigenvalues of  $A$ . Apply  $S$  to these equations to obtain for the  $n$ -vectors  $\vec{y}_i^j = S\vec{x}_i^j$  the Jordan chain relations

$$B\vec{y}_i^j = \lambda_i\vec{y}_i^j + \vec{y}_i^{j-1}, \quad j > 1, \quad B\vec{y}_i^1 = \lambda_i\vec{y}_i^1.$$

Because  $S$  has independent columns and the  $k$ -vectors  $\vec{x}_i^j$  are independent, then the  $n$ -vectors  $\vec{y}_i^j$  are independent.

The **plan** is to isolate the chains for eigenvalue zero, then extend these chains by one vector. Then 1-chains will be constructed from eigenpairs for eigenvalue zero to make  $n$  generalized eigenvectors.

Suppose  $q$  values of  $i$  satisfy  $\lambda_i = 0$ . We allow  $q = 0$ . For simplicity, assume such values  $i$  are  $i = 1, \dots, q$ . The key formula  $\vec{y}_i^j = S\vec{x}_i^j$  implies  $\vec{y}_i^j$  is in **Image**( $B$ ), while  $B\vec{y}_i^1 = \lambda_i\vec{y}_i^1$  implies  $\vec{y}_i^1, \dots, \vec{y}_i^q$  are in **kernel**( $B$ ). Each eigenvector  $\vec{y}_i^1$  starts a Jordan chain ending in  $\vec{y}_i^{m(i)}$ . Then<sup>1</sup> the equation  $B\vec{u} = \vec{y}_i^{m(i)}$  has an  $n$ -vector solution  $\vec{u}$ . We label  $\vec{u} = \vec{y}_i^{m(i)+1}$ . Because  $\lambda_i = 0$ , then  $B\vec{u} = \lambda_i\vec{u} + \vec{y}_i^{m(i)}$  results in an extended Jordan chain

$$\begin{aligned} B\vec{y}_i^1 &= \lambda_i\vec{y}_i^1 \\ B\vec{y}_i^2 &= \lambda_i\vec{y}_i^2 + \vec{y}_i^1 \\ &\vdots \\ B\vec{y}_i^{m(i)} &= \lambda_i\vec{y}_i^{m(i)} + \vec{y}_i^{m(i)-1} \\ B\vec{y}_i^{m(i)+1} &= \lambda_i\vec{y}_i^{m(i)+1} + \vec{y}_i^{m(i)} \end{aligned}$$

Let's extend the independent set  $\{\vec{y}_i^j\}_{i=1}^q$  to a basis of **kernel**( $B$ ) by adding  $s = n - k - q$  additional independent vectors  $\vec{v}_1, \dots, \vec{v}_s$ . This basis consists of eigenvectors of  $B$  for eigenvalue 0. Then the set of  $n$  vectors  $\vec{v}_r, \vec{y}_i^j$  for  $1 \leq r \leq s, 1 \leq i \leq q, 1 \leq j \leq m(i) + 1$  consists of eigenvectors of  $B$  and vectors that satisfy Jordan chain relations. These vectors are columns of a matrix  $\mathcal{P}$  that satisfies  $B\mathcal{P} = \mathcal{P}\mathcal{J}$  where  $\mathcal{J}$  is a Jordan form.

To prove  $\mathcal{P}$  invertible, assume a linear combination of the columns of  $\mathcal{P}$  is zero:

$$\sum_{i=q+1}^p \sum_{j=1}^{m(i)} b_i^j \vec{y}_i^j + \sum_{i=1}^q \sum_{j=1}^{m(i)+1} b_i^j \vec{y}_i^j + \sum_{i=1}^s c_i \vec{v}_i = \vec{0}.$$

Apply  $B$  to this equation. Because  $B\vec{w} = \vec{0}$  for any  $\vec{w}$  in **kernel**( $B$ ), then

$$\sum_{i=q+1}^p \sum_{j=1}^{m(i)} b_i^j B\vec{y}_i^j + \sum_{i=1}^q \sum_{j=2}^{m(i)+1} b_i^j B\vec{y}_i^j = \vec{0}.$$

The Jordan chain relations imply that the  $k$  vectors  $B\vec{y}_i^j$  in the linear combination consist of  $\lambda_i\vec{y}_i^j + \vec{y}_i^{j-1}$ ,  $\lambda_i\vec{y}_i^1$ ,  $i = q + 1, \dots, p, j = 2, \dots, m(i)$ , plus the vectors  $\vec{y}_i^j$ ,  $1 \leq i \leq q, 1 \leq j \leq m(i)$ . Independence of the original  $k$  vectors  $\{\vec{y}_i^j\}$  plus  $\lambda_i \neq 0$  for  $i > q$  implies this new set is independent. Then all coefficients in the linear combination are zero.

The first linear combination then reduces to  $\sum_{i=1}^q b_i^1 \vec{y}_i^1 + \sum_{i=1}^s c_i \vec{v}_i = \vec{0}$ . Independence of the constructed basis for **kernel**( $B$ ) implies  $b_i^1 = 0$  for  $1 \leq i \leq q$  and  $c_i = 0$  for  $1 \leq i \leq s$ . Therefore, the columns of  $\mathcal{P}$  are independent. The induction is complete.

<sup>1</sup>The  $n$ -vector  $\vec{u}$  is constructed by setting  $\vec{u} = \vec{0}$ , then copy components of  $k$ -vector  $\vec{x}_i^{m(i)}$  into pivot locations:  $\mathbf{row}(\vec{u}, i_j) = \mathbf{row}(\vec{x}_i^{m(i)}, j)$ ,  $j = 1, \dots, k$ .

**Geometric and algebraic multiplicity.** The **geometric multiplicity** is defined by  $\mathbf{GeoMult}(\lambda) = \dim(\mathbf{kernel}(A - \lambda I))$ , which is the number of basis vectors in a solution to  $(A - \lambda I)\vec{x} = \vec{0}$ , or, equivalently, the number of free variables. The **algebraic multiplicity** is the integer  $k = \mathbf{AlgMult}(\lambda)$  such that  $(r - \lambda)^k$  divides the characteristic polynomial  $\det(A - \lambda I)$ , but larger powers do not.

**Theorem 2 (Algebraic and Geometric Multiplicity)**

Let  $A$  be a square real or complex matrix. Then

$$(1) \quad 1 \leq \mathbf{GeoMult}(\lambda) \leq \mathbf{AlgMult}(\lambda).$$

In addition, there are the following relationships between the Jordan form  $J$  and algebraic and geometric multiplicities.

<b>GeoMult</b> ( $\lambda$ )	Equals the number of Jordan blocks in $J$ with eigenvalue $\lambda$ ,
<b>AlgMult</b> ( $\lambda$ )	Equals the number of times $\lambda$ is repeated along the diagonal of $J$ .

**Proof:** Let  $d = \mathbf{GeoMult}(\lambda_0)$ . Construct a basis  $v_1, \dots, v_n$  of  $\mathcal{R}^n$  such that  $v_1, \dots, v_d$  is a basis for  $\mathbf{kernel}(A - \lambda_0 I)$ . Define  $S = \langle v_1 | \dots | v_n \rangle$  and  $B = S^{-1}AS$ . The first  $d$  columns of  $AS$  are  $\lambda_0 v_1, \dots, \lambda_0 v_d$ . Then  $B = \left( \begin{array}{c|c} \lambda_0 I & C \\ \hline 0 & D \end{array} \right)$  for some matrices  $C$  and  $D$ . Cofactor expansion implies some polynomial  $g$  satisfies

$$\det(A - \lambda I) = \det(S(B - \lambda I)S^{-1}) = \det(B - \lambda I) = (\lambda - \lambda_0)^d g(\lambda)$$

and therefore  $d \leq \mathbf{AlgMult}(\lambda_0)$ . Other details of proof are left to the reader.

**Chains of generalized eigenvectors.** Given an eigenvalue  $\lambda$  of the matrix  $A$ , the topic of generalized eigenanalysis determines a Jordan block  $B(\lambda, m)$  in  $J$  by finding an  $m$ -**chain** of generalized eigenvectors  $\vec{v}_1, \dots, \vec{v}_m$ , which appear as columns of  $P$  in the relation  $A = PJP^{-1}$ . The very first vector  $\vec{v}_1$  of the chain is an eigenvector,  $(A - \lambda I)\vec{v}_1 = \vec{0}$ . The others  $\vec{v}_2, \dots, \vec{v}_k$  are not eigenvectors but satisfy

$$(A - \lambda I)\vec{v}_2 = \vec{v}_1, \quad \dots \quad , \quad (A - \lambda I)\vec{v}_m = \vec{v}_{m-1}.$$

Implied by the term  $m$ -**chain** is insolvability of  $(A - \lambda I)\vec{x} = \vec{v}_m$ . The chain size  $m$  is subject to the inequality  $1 \leq m \leq \mathbf{AlgMult}(\lambda)$ .

The Jordan form  $J$  may contain several Jordan blocks for one eigenvalue  $\lambda$ . To illustrate, if  $J$  has only one eigenvalue  $\lambda$  and  $\mathbf{AlgMult}(\lambda) = 3$ ,

then  $J$  might be constructed as follows:

$$\begin{aligned} J = \mathbf{diag}(B(\lambda, 1), B(\lambda, 1), B(\lambda, 1)) &= \begin{pmatrix} \lambda & 0 & 0 \\ 0 & \lambda & 0 \\ 0 & 0 & \lambda \end{pmatrix}, \\ J = \mathbf{diag}(B(\lambda, 1), B(\lambda, 2)) &= \begin{pmatrix} \lambda & 0 & 0 \\ 0 & \lambda & 1 \\ 0 & 0 & \lambda \end{pmatrix}, \\ J = B(\lambda, 3) &= \begin{pmatrix} \lambda & 1 & 0 \\ 0 & \lambda & 1 \\ 0 & 0 & \lambda \end{pmatrix}. \end{aligned}$$

The three generalized eigenvectors for this example correspond to

$$\begin{aligned} J = \begin{pmatrix} \lambda & 0 & 0 \\ 0 & \lambda & 0 \\ 0 & 0 & \lambda \end{pmatrix} &\leftrightarrow \text{Three 1-chains,} \\ J = \begin{pmatrix} \lambda & 0 & 0 \\ 0 & \lambda & 1 \\ 0 & 0 & \lambda \end{pmatrix} &\leftrightarrow \text{One 1-chain and one 2-chain,} \\ J = \begin{pmatrix} \lambda & 1 & 0 \\ 0 & \lambda & 1 \\ 0 & 0 & \lambda \end{pmatrix} &\leftrightarrow \text{One 3-chain.} \end{aligned}$$

**Computing  $m$ -chains.** Let us fix the discussion to an eigenvalue  $\lambda$  of  $A$ . Define  $N = A - \lambda I$  and  $p = \mathbf{AlgMult}(\lambda)$ .

To compute an  $m$ -chain, start with an eigenvector  $\vec{v}_1$  and solve recursively by **rref** methods  $N\vec{v}_{j+1} = \vec{v}_j$  until there fails to be a solution. This must seemingly be done for *all possible choices* of  $\vec{v}_1$ ! The search for  $m$ -chains terminates when  $p$  independent generalized eigenvectors have been calculated.

If  $A$  has an essentially unique eigenpair  $(\lambda, \vec{v}_1)$ , then this process terminates immediately with an  $m$ -chain where  $m = p$ . The chain produces one Jordan block  $B(\lambda, m)$  and the generalized eigenvectors  $\vec{v}_1, \dots, \vec{v}_m$  are recorded into the matrix  $P$ .

If  $\vec{u}_1, \vec{u}_2$  form a basis for the eigenvectors of  $A$  corresponding to  $\lambda$ , then the problem  $N\vec{x} = \vec{0}$  has 2 free variables. Therefore, we seek to find an  $m_1$ -chain and an  $m_2$ -chain such that  $m_1 + m_2 = p$ , corresponding to two Jordan blocks  $B(\lambda, m_1)$  and  $B(\lambda, m_2)$ .

To understand the logic applied here, the reader should verify that for  $\mathcal{N} = \mathbf{diag}(B(0, m_1), B(0, m_2), \dots, B(0, m_k))$  the problem  $\mathcal{N}\vec{x} = \vec{0}$  has  $k$  free variables, because  $\mathcal{N}$  is already in **rref** form. These remarks imply that a  $k$ -dimensional basis of eigenvectors of  $A$  for eigenvalue  $\lambda$

causes a search for  $m_i$ -chains,  $1 \leq i \leq k$ , such that  $m_1 + \cdots + m_k = p$ , corresponding to  $k$  Jordan blocks  $B(\lambda, m_1), \dots, B(\lambda, m_k)$ .

A common naive approach for computing generalized eigenvectors can be illustrated by letting

$$A = \begin{pmatrix} 1 & 1 & 1 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}, \quad \vec{\mathbf{u}}_1 = \begin{pmatrix} 1 \\ -1 \\ 1 \end{pmatrix}, \quad \vec{\mathbf{u}}_2 = \begin{pmatrix} 0 \\ 1 \\ -1 \end{pmatrix}.$$

Matrix  $A$  has one eigenvalue  $\lambda = 1$  and two eigenpairs  $(1, \vec{\mathbf{u}}_1), (1, \vec{\mathbf{u}}_2)$ . Starting a chain calculation with  $\vec{\mathbf{v}}_1$  equal to either  $\vec{\mathbf{u}}_1$  or  $\vec{\mathbf{u}}_2$  gives a 1-chain. This naive approach leads to only two independent generalized eigenvectors. However, the calculation must proceed until three independent generalized eigenvectors have been computed. To resolve the trouble, keep a 1-chain, say the one generated by  $\vec{\mathbf{u}}_1$ , and start a new chain calculation using  $\vec{\mathbf{v}}_1 = a_1\vec{\mathbf{u}}_1 + a_2\vec{\mathbf{u}}_2$ . Adjust the values of  $a_1, a_2$  until a 2-chain has been computed:

$$\langle A - \lambda I | \vec{\mathbf{v}}_1 \rangle = \begin{pmatrix} 0 & 1 & 1 & a_1 \\ 0 & 0 & 0 & -a_1 + a_2 \\ 0 & 0 & 0 & a_1 - a_2 \end{pmatrix} \approx \begin{pmatrix} 0 & 1 & 1 & a_1 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix},$$

provided  $a_1 - a_2 = 0$ . Choose  $a_1 = a_2 = 1$  to make  $\vec{\mathbf{v}}_1 = \vec{\mathbf{u}}_1 + \vec{\mathbf{u}}_2 \neq \vec{\mathbf{0}}$  and solve for  $\vec{\mathbf{v}}_2 = (0, 1, 0)$ . Then  $\vec{\mathbf{u}}_1$  is a 1-chain and  $\vec{\mathbf{v}}_1, \vec{\mathbf{v}}_2$  is a 2-chain. The generalized eigenvectors  $\vec{\mathbf{u}}_1, \vec{\mathbf{v}}_1, \vec{\mathbf{v}}_2$  are independent and form the columns of  $P$  while  $J = \mathbf{diag}(B(\lambda, 1), B(\lambda, 2))$  (recall  $\lambda = 1$ ). We justify  $A = PJP^{-1}$  by testing  $AP = PJ$ , using the formulas

$$J = \begin{pmatrix} \lambda & 0 & 0 \\ 0 & \lambda & 1 \\ 0 & 0 & \lambda \end{pmatrix}, \quad P = \begin{pmatrix} 1 & 1 & 0 \\ -1 & 0 & 1 \\ 1 & 0 & 0 \end{pmatrix}.$$

## Jordan Decomposition using maple

Displayed here is `maple` code which applied to the matrix

$$A = \begin{pmatrix} 4 & -2 & 5 \\ -2 & 4 & -3 \\ 0 & 0 & 2 \end{pmatrix}$$

produces the Jordan decomposition

$$A = PJP^{-1} = \frac{1}{4} \begin{pmatrix} 1 & 4 & -7 \\ -1 & 4 & 1 \\ 0 & 0 & 4 \end{pmatrix} \begin{pmatrix} 6 & 0 & 0 \\ 0 & 2 & 1 \\ 0 & 0 & 2 \end{pmatrix} \frac{1}{4} \begin{pmatrix} 8 & -8 & 16 \\ 2 & 2 & 3 \\ 0 & 0 & 4 \end{pmatrix}.$$

```

A := Matrix([[4, -2, 5], [-2, 4, -3], [0, 0, 2]]);
factor(LinearAlgebra[CharacteristicPolynomial](A,lambda));
# Answer == (lambda-6)*(lambda-2)^2
J,P:=LinearAlgebra[JordanForm](A,output=['J','Q']);
zero:=A.P-P.J; # zero matrix expected

```

## Number of Jordan Blocks

In calculating generalized eigenvectors of  $A$  for eigenvalue  $\lambda$ , it is possible to decide in advance how many Jordan chains of size  $k$  should be computed. A practical consequence is to organize the computation for certain chain sizes.

### Theorem 3 (Number of Jordan Blocks)

Given eigenvalue  $\lambda$  of  $A$ , define  $N = A - \lambda I$ ,  $k(j) = \dim(\mathbf{kernel}(N^j))$ . Let  $p$  be the least integer such that  $N^p = N^{p+1}$ . Then the Jordan form of  $A$  has  $2k(j-1) - k(j-2) - k(j)$  Jordan blocks  $B(\lambda, j-1)$ ,  $j = 3, \dots, p$ .

The proof of the theorem is in the exercises, where more detail appears for  $p = 1$  and  $p = 2$ . Complete results are in the `maple` code below.

**An Illustration.** This example is a  $5 \times 5$  matrix  $A$  with one eigenvalue  $\lambda = 2$  of multiplicity 5. Let  $s(j) =$  number of  $j \times j$  Jordan blocks.

$$A = \begin{pmatrix} 3 & -1 & 1 & 0 & 0 \\ 2 & 0 & 1 & 1 & 0 \\ 1 & -1 & 2 & 1 & 0 \\ -1 & 1 & 0 & 2 & 1 \\ -3 & 3 & 0 & -2 & 3 \end{pmatrix}, \quad N = A - 2I = \begin{pmatrix} 1 & -1 & 1 & 0 & 0 \\ 2 & -2 & 1 & 1 & 0 \\ 1 & -1 & 0 & 1 & 0 \\ -1 & 1 & 0 & 0 & 1 \\ -3 & 3 & 0 & -2 & 1 \end{pmatrix}.$$

Then  $N^3 = N^4 = N^5 = 0$  implies  $k(3) = k(4) = k(5) = 5$ . Further,  $k(2) = 4$ ,  $k(1) = 2$ . Then  $s(5) = s(4) = 0$ ,  $s(3) = s(2) = 1$ ,  $s(1) = 0$ , which implies one block of each size 2 and 3.

Some `maple` code automates the investigation:

```

with(LinearAlgebra):
A := Matrix([
[ 3, -1, 1, 0, 0],[ 2, 0, 1, 1, 0],
[ 1, -1, 2, 1, 0],[-1, 1, 0, 2, 1],
[-3, 3, 0, -2, 3] ]);
lambda:=2;
n:=RowDimension(A);N:=A-lambda*IdentityMatrix(n);
for j from 1 to n do
  k[j]:=n-Rank(N^j); od:
for p from n to 2 by -1 do

```

```

if(k[p]<>k[p-1])then break; fi: od;
txt:=(j,x)->printf('if'(x=1,
  cat("B(lambda,"j,") occurs 1 time\n"),
  cat("B(lambda,"j,") occurs ",x," times\n"))):
printf("lambda=%d, nilpotency=%d\n",lambda,p);
if(p=1) then txt(1,k[1]); else
  txt(p,k[p]-k[p-1]);
  for j from p to 3 by -1 do
    txt(j-1,2*k[j-1]-k[j-2]-k[j]): od:
  txt(1,2*k[1]-k[2]);
fi:
#lambda=2, nilpotency=3
#B(lambda,3) occurs 1 time
#B(lambda,2) occurs 1 time
#B(lambda,1) occurs 0 times
J,P:=JordanForm(A,output=['J','Q']):
# Answer check for the maple code

```

$$J = \begin{pmatrix} 2 & 1 & 0 & 0 & 0 \\ 0 & 2 & 1 & 0 & 0 \\ 0 & 0 & 2 & 0 & 0 \\ 0 & 0 & 0 & 2 & 1 \\ 0 & 0 & 0 & 0 & 2 \end{pmatrix}, \quad P = \frac{1}{2} \begin{pmatrix} 0 & 1 & 2 & -1 & 0 \\ -4 & 2 & 2 & -2 & 2 \\ -4 & 1 & 1 & -1 & 1 \\ -4 & -3 & 1 & -1 & 1 \\ 4 & -5 & -3 & 1 & -3 \end{pmatrix}$$

## Numerical Instability

The matrix  $A = \begin{pmatrix} 1 & 1 \\ \varepsilon & 1 \end{pmatrix}$  has two possible Jordan forms

$$J(\varepsilon) = \begin{cases} \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix} & \varepsilon = 0, \\ \begin{pmatrix} 1 + \sqrt{\varepsilon} & 0 \\ 0 & 1 - \sqrt{\varepsilon} \end{pmatrix} & \varepsilon > 0. \end{cases}$$

When  $\varepsilon \approx 0$ , then numerical algorithms become unstable, unable to lock onto the correct Jordan form. Briefly,  $\lim_{\varepsilon \rightarrow 0} J(\varepsilon) \neq J(0)$ .

## The Real Jordan Form of $A$

Given a real matrix  $A$ , generalized eigenanalysis seeks to find a *real* invertible matrix  $\mathcal{P}$  and a *real* upper triangular block matrix  $R$  such that  $A = \mathcal{P}R\mathcal{P}^{-1}$ .

If  $\lambda$  is a real eigenvalue of  $A$ , then a **real Jordan block** is a matrix

$$B = \mathbf{diag}(\lambda, \dots, \lambda) + N, \quad N = \begin{pmatrix} 0 & 1 & 0 & \cdots & 0 \\ \vdots & \vdots & \vdots & \cdots & \vdots \\ 0 & 0 & 0 & \cdots & 1 \\ 0 & 0 & 0 & \cdots & 0 \end{pmatrix}.$$

If  $\lambda = a + ib$  is a complex eigenvalue of  $A$ , then symbols  $\lambda$ , 1 and 0 are replaced respectively by  $2 \times 2$  real matrices  $\Lambda = \begin{pmatrix} a & b \\ -b & a \end{pmatrix}$ ,  $\mathcal{I} = \mathbf{diag}(1, 1)$  and  $\mathcal{O} = \mathbf{diag}(0, 0)$ . The corresponding  $2m \times 2m$  real Jordan block matrix is given by the formula

$$B = \mathbf{diag}(\Lambda, \dots, \Lambda) + \mathcal{N}, \quad \mathcal{N} = \begin{pmatrix} \mathcal{O} & \mathcal{I} & \mathcal{O} & \cdots & \mathcal{O} & \mathcal{O} \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ \mathcal{O} & \mathcal{O} & \mathcal{O} & \cdots & \mathcal{O} & \mathcal{I} \\ \mathcal{O} & \mathcal{O} & \mathcal{O} & \cdots & \mathcal{O} & \mathcal{O} \end{pmatrix}.$$

## Direct Sum Decomposition

The **generalized eigenspace** of eigenvalue  $\lambda$  of an  $n \times n$  matrix  $A$  is the subspace  $\mathbf{kernel}((A - \lambda I)^p)$  where  $p = \mathbf{AlgMult}(\lambda)$ . We state without proof the main result for generalized eigenspace bases, because details appear in the exercises. A proof is included for the direct sum decomposition, even though Putzer's spectral theory independently produces the same decomposition.

### Theorem 4 (Generalized Eigenspace Basis)

The subspace  $\mathbf{kernel}((A - \lambda I)^k)$ ,  $k = \mathbf{AlgMult}(\lambda)$  has a  $k$ -dimensional basis whose vectors are the columns of  $P$  corresponding to blocks  $B(\lambda, j)$  of  $J$ , in Jordan decomposition  $A = PJP^{-1}$ .

### Theorem 5 (Direct Sum Decomposition)

Given  $n \times n$  matrix  $A$  and distinct eigenvalues  $\lambda_1, \dots, \lambda_k$ ,  $n_1 = \mathbf{AlgMult}(\lambda_1)$ ,  $\dots$ ,  $n_k = \mathbf{AlgMult}(\lambda_k)$ , then  $A$  induces a direct sum decomposition

$$\mathcal{C}^n = \mathbf{kernel}((A - \lambda_1 I)^{n_1}) \oplus \cdots \oplus \mathbf{kernel}((A - \lambda_k I)^{n_k}).$$

This equation means that each complex vector  $\vec{x}$  in  $\mathcal{C}^n$  can be uniquely written as

$$\vec{x} = \vec{x}_1 + \cdots + \vec{x}_k$$

where each  $\vec{x}_i$  belongs to  $\mathbf{kernel}((A - \lambda_i I)^{n_i})$ ,  $i = 1, \dots, k$ .

**Proof:** The previous theorem implies there is a basis of dimension  $n_i$  for  $E_i \equiv \mathbf{kernel}((A - \lambda_i I)^{n_i})$ ,  $i = 1, \dots, k$ . Because  $n_1 + \cdots + n_k = n$ , then there are  $n$  vectors in the union of these bases. The independence test for these  $n$  vectors

amounts to showing that  $\vec{x}_1 + \cdots + \vec{x}_k = \vec{0}$  with  $\vec{x}_i$  in  $E_i$ ,  $i = 1, \dots, k$ , implies all  $\vec{x}_i = \vec{0}$ . This will be true provided  $E_i \cap E_j = \{\vec{0}\}$  for  $i \neq j$ .

Let's assume a Jordan decomposition  $A = PJP^{-1}$ . If  $\vec{x}$  is common to both  $E_i$  and  $E_j$ , then basis expansion of  $\vec{x}$  in both subspaces implies a linear combination of the columns of  $P$  is zero, which by independence of the columns of  $P$  implies  $\vec{x} = \vec{0}$ .

The proof is complete.

## Computing Exponential Matrices

Discussed here are methods for finding a real exponential matrix  $e^{At}$  when  $A$  is real. Two formulas are given, one for a real eigenvalue and one for a complex eigenvalue. These formulas supplement the spectral formulas given earlier in the text.

**Nilpotent matrices.** A matrix  $N$  which satisfies  $N^p = 0$  for some integer  $p$  is called **nilpotent**. The least integer  $p$  for which  $N^p = 0$  is called the **nilpotency** of  $N$ . A nilpotent matrix  $N$  has a finite exponential series:

$$e^{Nt} = I + Nt + N^2 \frac{t^2}{2!} + \cdots + N^{p-1} \frac{t^{p-1}}{(p-1)!}.$$

If  $N = B(\lambda, p) - \lambda I$ , then the finite sum has a splendidly simple expression. Due to  $e^{\lambda t + Nt} = e^{\lambda t} e^{Nt}$ , this proves the following result.

### Theorem 6 (Exponential of a Jordan Block Matrix)

If  $\lambda$  is real and

$$B = \begin{pmatrix} \lambda & 1 & 0 & \cdots & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & \cdots & \lambda & 1 \\ 0 & 0 & 0 & \cdots & 0 & \lambda \end{pmatrix} \quad (p \times p \text{ matrix})$$

then

$$e^{Bt} = e^{\lambda t} \begin{pmatrix} 1 & t & \frac{t^2}{2} & \cdots & \frac{t^{p-2}}{(p-2)!} & \frac{t^{p-1}}{(p-1)!} \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & \cdots & 1 & t \\ 0 & 0 & 0 & \cdots & 0 & 1 \end{pmatrix}.$$

The equality also holds if  $\lambda$  is a complex number, in which case both sides of the equation are complex.

**Real Exponentials for Complex  $\lambda$ .** A Jordan decomposition  $A = \mathcal{P}J\mathcal{P}^{-1}$ , in which  $A$  has only real eigenvalues, has real generalized eigenvectors appearing as columns in the matrix  $\mathcal{P}$ , in the natural order given in  $J$ . When  $\lambda = a + ib$  is complex,  $b > 0$ , then the real and imaginary parts of each generalized eigenvector are entered pairwise into  $\mathcal{P}$ ; the conjugate eigenvalue  $\bar{\lambda} = a - ib$  is skipped. The complex entry along the diagonal of  $J$  is changed into a  $2 \times 2$  matrix under the correspondence

$$a + ib \leftrightarrow \begin{pmatrix} a & b \\ -b & a \end{pmatrix}.$$

The result is a *real* matrix  $\mathcal{P}$  and a *real* block upper triangular matrix  $J$  which satisfy  $A = \mathcal{P}J\mathcal{P}^{-1}$ .

**Theorem 7 (Real Block Diagonal Matrix, Eigenvalue  $a + ib$ )**

Let  $\Lambda = \begin{pmatrix} a & b \\ -b & a \end{pmatrix}$ ,  $\mathcal{I} = \mathbf{diag}(1, 1)$  and  $\mathcal{O} = \mathbf{diag}(0, 0)$ . Consider a real Jordan block matrix of dimension  $2m \times 2m$  given by the formula

$$B = \begin{pmatrix} \Lambda & \mathcal{I} & \mathcal{O} & \cdots & \mathcal{O} & \mathcal{O} \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ \mathcal{O} & \mathcal{O} & \mathcal{O} & \cdots & \Lambda & \mathcal{I} \\ \mathcal{O} & \mathcal{O} & \mathcal{O} & \cdots & \mathcal{O} & \Lambda \end{pmatrix}.$$

If  $\mathcal{R} = \begin{pmatrix} \cos bt & \sin bt \\ -\sin bt & \cos bt \end{pmatrix}$ , then

$$e^{Bt} = e^{at} \begin{pmatrix} \mathcal{R} & t\mathcal{R} & \frac{t^2}{2}\mathcal{R} & \cdots & \frac{t^{m-2}}{(m-2)!}\mathcal{R} & \frac{t^{m-1}}{(m-1)!}\mathcal{R} \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ \mathcal{O} & \mathcal{O} & \mathcal{O} & \cdots & \mathcal{R} & t\mathcal{R} \\ \mathcal{O} & \mathcal{O} & \mathcal{O} & \cdots & \mathcal{O} & \mathcal{R} \end{pmatrix}.$$

**Solving  $\vec{x}' = A\vec{x}$ .** The solution  $\vec{x}(t) = e^{At}\vec{x}(0)$  must be real if  $A$  is real. The real solution can be expressed as  $\vec{x}(t) = \mathcal{P}\vec{y}(t)$  where  $\vec{y}'(t) = R\vec{y}(t)$  and  $R$  is a real Jordan form of  $A$ , containing real Jordan blocks  $B_1, \dots, B_k$  down its diagonal. Theorems above provide explicit formulas for the block matrices  $e^{B_i t}$  in the relation

$$e^{Rt} = \mathbf{diag}(e^{B_1 t}, \dots, e^{B_k t}).$$

The resulting formula

$$\vec{x}(t) = \mathcal{P}e^{Rt}\mathcal{P}^{-1}\vec{x}(0)$$

contains only real numbers, real exponentials, plus sine and cosine terms, which are possibly multiplied by polynomials in  $t$ .

## Exercises 10.1

Jordan block. Write out explicitly.

- 1.
- 2.
- 3.
- 4.

Jordan form. Which are Jordan forms and which are not? Explain.

- 5.
- 6.
- 7.
- 8.

Decoding  $A = PJP^{-1}$ . Decode  $A = PJP^{-1}$  in each case, displaying explicitly the Jordan chain relations.

- 9.
- 10.
- 11.
- 12.

Geometric multiplicity. Determine the geometric multiplicity  $\mathbf{GeoMult}(\lambda)$ .

- 13.
- 14.
- 15.
- 16.

Algebraic multiplicity. Determine the algebraic multiplicity  $\mathbf{AlgMult}(\lambda)$ .

- 17.
- 18.
- 19.

20.

Generalized eigenvectors. Find all generalized eigenvectors and represent  $A = PJP^{-1}$ .

- 21.
- 22.
- 23.
- 24.
- 25.
- 26.
- 27.
- 28.
- 29.

- 30.
- 31.
- 32.

Computing  $m$ -chains. Find the Jordan chains for the given eigenvalue.

- 33.
- 34.
- 35.
- 36.
- 37.
- 38.
- 39.
- 40.

Jordan Decomposition. Use `maple` to find the Jordan decomposition.

- 41.
- 42.
- 43.

- 44.
- 45.
- 46.
- 47.
- 48.

Number of Jordan Blocks. Outlined here is the derivation of

$$s(j) = 2k(j - 1) - k(j - 2) - k(j).$$

Definitions:

- $s(j)$  = number of blocks  $B(\lambda, j)$
- $N = A - \lambda I$
- $k(j) = \dim(\mathbf{kernel}(N^j))$
- $L_j = \mathbf{kernel}(N^{j-1})^\perp$  relative to  $\mathbf{kernel}(N^j)$
- $\ell(j) = \dim(L_j)$
- $p$  minimizes  $\mathbf{kernel}(N^p) = \mathbf{kernel}(N^{p+1})$

- 49. Verify  $k(j) \leq k(j + 1)$  from

$$\mathbf{kernel}(N^j) \subset \mathbf{kernel}(N^{j+1}).$$

- 50. Verify the direct sum formula

$$\mathbf{kernel}(N^j) = \mathbf{kernel}(N^{j-1}) \oplus L_j.$$

Then  $k(j) = k(j - 1) + \ell(j)$ .

- 51. Given  $N^j \vec{v} = \vec{0}$ ,  $N^{j-1} \vec{v} \neq \vec{0}$ , define  $\vec{v}_i = N^{j-i} \vec{v}$ ,  $i = 1, \dots, j$ . Show that these are independent vectors satisfying Jordan chain relations  $N \vec{v}_1 = \vec{0}$ ,  $N \vec{v}_{i+1} = \vec{v}_i$ .
- 52. A block  $B(\lambda, p)$  corresponds to a Jordan chain  $\vec{v}_1, \dots, \vec{v}_p$  constructed from the Jordan decomposition. Use  $N^{j-1} \vec{v}_j = \vec{v}_1$  and  $\mathbf{kernel}(N^p) = \mathbf{kernel}(N^{p+1})$  to show that the number of such blocks  $B(\lambda, p)$  is  $\ell(p)$ . Then for  $p > 1$ ,  $s(p) = k(p) - k(p - 1)$ .

- 53. Show that  $\ell(j - 1) - \ell(j)$  is the number of blocks  $B(\lambda, j)$  for  $2 < j < p$ . Then

$$s(j) = 2k(j - 1) - k(j) - k(j - 2).$$

- 54. Test the formulas above on the special matrices

$$A = \mathbf{diag}(B(\lambda, 1), B(\lambda, 1), B(\lambda, 1)),$$

$$A = \mathbf{diag}(B(\lambda, 1), B(\lambda, 2), B(\lambda, 3)),$$

$$A = \mathbf{diag}(B(\lambda, 1), B(\lambda, 3), B(\lambda, 3)),$$

$$A = \mathbf{diag}(B(\lambda, 1), B(\lambda, 1), B(\lambda, 3)),$$

Generalized Eigenspace Basis.

Let  $A$  be  $n \times n$  with distinct eigenvalues  $\lambda_i$ ,  $n_i = \mathbf{AlgMult}(\lambda_i)$  and  $E_i = \mathbf{kernel}((A - \lambda_i I)^{n_i})$ ,  $i = 1, \dots, k$ . Assume a Jordan decomposition  $A = PJP^{-1}$ .

- 55. Let Jordan block  $B(\lambda, j)$  appear in  $J$ . Prove that a Jordan chain corresponding to this block is a set of  $j$  independent columns of  $P$ .

- 56. Let  $\mathcal{B}_\lambda$  be the union of all columns of  $P$  originating from Jordan chains associated with Jordan blocks  $B(\lambda, j)$ . Prove that  $\mathcal{B}_\lambda$  is an independent set.

- 57. Verify that  $\mathcal{B}_\lambda$  has  $\mathbf{AlgMult}(\lambda)$  basis elements.

- 58. Prove that  $E_i = \mathbf{span}(\mathcal{B}_{\lambda_i})$  and  $\dim(E_i) = n_i$ ,  $i = 1, \dots, k$ .

Numerical Instability. Show directly that  $\lim_{\epsilon \rightarrow 0} J(\epsilon) \neq J(0)$ .

- 59.
- 60.
- 61.
- 62.
- 63.
- 64.

Direct Sum Decomposition. Display the direct sum decomposition.

65.

66.

67.

68.

69.

70.

Exponential Matrices. Compute the exponential matrix on paper and then check the answer using `maple`.

71.

72.

73.

74.

75.

76.

77.

78.

Nilpotent matrices. Find the nilpotency of  $N$ .

79.

80.

81.

82.

Real Exponentials. Compute the real exponential  $e^{At}$  on paper. Check the answer in `maple`.

83.

84.

85.

86.

Real Jordan Form. Find the real Jordan form.

87.

88.

89.

90.

Solving  $\vec{x}' = A\vec{x}$ . Solve the differential equation.

91.

92.

93.

94.