

## 1 Solving Linear Recurrences

It is straightforward to verify that the recurrence

$$\mathbf{x}_{n+1} = A\mathbf{x}_n + \mathbf{f}_n, \quad \mathbf{x}_0 \text{ given} \quad (1)$$

where  $\mathbf{x}_n$  is a  $m$ -vector and  $A$  a  $m \times m$  constant matrix, has solution

$$\mathbf{x}_n = A^n \mathbf{x}_0 + \sum_{j=0}^{n-1} A^{n-1-j} \mathbf{f}_j \quad (2)$$

As it stands, this formula gives limited information about the nature of  $A^n$ . To get a better handle on this, we need to consider the eigenstructure of the matrix.

## 2 Eigenvalues and Eigenvectors

Let  $A$  be a  $m \times m$  matrix whose elements are members of the field  $K$ , ( $K = \mathbf{R}$  or  $\mathbf{C}$ ),  $\lambda \in K$  and  $\mathbf{e} \neq \mathbf{0}$  a  $m$ -vector such that

$$A\mathbf{e} = \lambda\mathbf{e} \quad (3)$$

then  $\lambda$  is an *eigenvalue* of  $A$ , and  $\mathbf{e}$  a corresponding eigenvector.

From Eq(3)

$$(\lambda I - A)\mathbf{e} = \mathbf{0}$$

and so  $\lambda$  must satisfy the CHARACTERISTIC equation:

$$\det(\lambda I - A) = 0 \quad (4)$$

It can be shown that

$$\det(\lambda I - A) = \lambda^m + a_{m-1}\lambda^{m-1} + \dots + a_1\lambda + a_0 = (\lambda - \lambda_1)(\lambda - \lambda_2) \cdots (\lambda - \lambda_m) \quad (5)$$

where  $\lambda_1, \lambda_2, \dots, \lambda_m$  are the eigenvalues of  $A$ . This is called the characteristic polynomial of  $A$ .

Note: In theory, Eq(4) can be used to find  $\lambda$ 's – than Eq(3) used to find the corresponding  $\mathbf{e}$ 's.

Example:

$$A = \begin{pmatrix} 0 & 1 \\ -2 & -3 \end{pmatrix}$$

$$\begin{aligned} \det(\lambda I - A) &= \det \begin{pmatrix} \lambda & -1 \\ 2 & \lambda + 3 \end{pmatrix} \\ &= \lambda^2 + 3\lambda + 2 \\ &= 0 \Rightarrow \lambda = -1 \text{ or } -2 \end{aligned}$$

For  $\lambda_1 = -1$

$$\begin{pmatrix} 0 & 1 \\ -2 & -3 \end{pmatrix} \mathbf{e}_1 = (-1)\mathbf{e}_1 \Rightarrow \mathbf{e}_1 = \alpha_1 \begin{pmatrix} 1 \\ -1 \end{pmatrix}, \quad \alpha_1 \in K$$

For  $\lambda_2 = -2$

$$\begin{pmatrix} 0 & 1 \\ -2 & -3 \end{pmatrix} \mathbf{e}_2 = (-2)\mathbf{e}_2 \Rightarrow \mathbf{e}_2 = \alpha_2 \begin{pmatrix} 1 \\ -2 \end{pmatrix}, \quad \alpha_2 \in K$$

### 3 Similarity

Two  $m \times m$  matrices  $R$  and  $S$  are said to be *similar* if there exist an *invertible* matrix  $P$  such that

$$R = P^{-1}SP$$

Similar matrices have the same *spectrum* (set of eigenvalues):

$$\begin{aligned} \det(\lambda I - R) &= \det(\lambda I - P^{-1}SP) \\ &= \det(\lambda P^{-1}P - P^{-1}SP) \\ &= \det(P^{-1}(\lambda I - S)P) \\ &= \det P^{-1} \det(\lambda I - S) \det P \\ &= \det(\lambda I - S) \end{aligned}$$

### 4 Diagonalisability

Recall that a diagonal matrix is a  $m \times m$  matrix, all of whose off-diagonal entries are zero. We denote a diagonal matrix by  $\text{diag}\{d_1, d_2, \dots, d_m\}$  where  $d_1, d_2, \dots, d_m$  are the diagonal entries.

The  $m \times m$  matrix  $A$  is said to be *diagonalisable* if it is similar to a diagonal matrix. It can be shown that the diagonal entries of the diagonal matrix are the eigenvalues of  $A$ .

Not every square matrix is diagonalisable. A necessary and sufficient condition for  $A$  to be diagonalisable is that its eigenvectors form a *linearly independent* set. Let  $A$  have spectrum  $\lambda_1, \lambda_2, \dots, \lambda_m$  with associated eigenvectors  $\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_m$  respectively. Then, for  $i = 1, 2, \dots, m$  we have  $A\mathbf{e}_i = \lambda_i\mathbf{e}_i$ . Consider

$$\begin{aligned} A E = A [\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_m] &= [A\mathbf{e}_1, A\mathbf{e}_2, \dots, A\mathbf{e}_m] \\ &= [\lambda_1\mathbf{e}_1, \lambda_2\mathbf{e}_2, \dots, \lambda_m\mathbf{e}_m] \\ &= [\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_m] \text{diag}\{\lambda_1, \lambda_2, \dots, \lambda_m\} \\ &= E \Lambda \end{aligned}$$

where  $E = [\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_m]$  and  $\Lambda = \text{diag}\{\lambda_1, \lambda_2, \dots, \lambda_m\}$ .

Thus, if  $A$  is diagonalisable

$$A = E \Lambda E^{-1} \tag{6}$$

$\Lambda$  is unique up to ordering of the eigenvalues.

We note that it can be shown that eigenvectors corresponding to distinct eigenvalues are linearly independent; hence, if  $A$  has  $m$  distinct eigenvalues, it is diagonalisable.

Example (Cont'd):  $A = E \Lambda E^{-1}$

$$\begin{pmatrix} 0 & 1 \\ -2 & -3 \end{pmatrix} = \begin{pmatrix} 1 & 1 \\ -1 & -2 \end{pmatrix} \begin{pmatrix} -1 & 0 \\ 0 & -2 \end{pmatrix} \begin{pmatrix} 2 & 1 \\ -1 & -1 \end{pmatrix}$$

We also note that the following corollary of Eq(6) which may be proven using induction

$$A^n = E \Lambda^n E^{-1} \quad n = 0, 1, \dots \tag{7}$$

where it is also straightforward to show that

$$\Lambda^n = \text{diag}\{\lambda_1^n, \lambda_2^n, \dots, \lambda_m^n\} \tag{8}$$

Example(Cont'd):  $A^n = E \Lambda^n E^{-1}$

$$\begin{aligned} \begin{pmatrix} 0 & 1 \\ -2 & -3 \end{pmatrix}^n &= \begin{pmatrix} 1 & 1 \\ -1 & -2 \end{pmatrix} \begin{pmatrix} (-1)^n & 0 \\ 0 & (-2)^n \end{pmatrix} \begin{pmatrix} 2 & 1 \\ -1 & -1 \end{pmatrix} \\ &= \begin{pmatrix} 2(-1)^n - (-2)^n & (-1)^n - (-2)^n \\ -2(-1)^n + 2(-2)^n & -(-1)^n + 2(-2)^n \end{pmatrix} \end{aligned}$$

Thus the solution of

$$\mathbf{x}_{n+1} = \begin{pmatrix} 0 & 1 \\ -2 & -3 \end{pmatrix} \mathbf{x}_n, \quad \mathbf{x}_0 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

is given by

$$\begin{aligned} \mathbf{x}_n &= \begin{pmatrix} 0 & 1 \\ -2 & -3 \end{pmatrix}^n \begin{pmatrix} 1 \\ 0 \end{pmatrix} \\ &= \begin{pmatrix} 2(-1)^n - (-2)^n \\ -2(-1)^n + 2(-2)^n \end{pmatrix} \end{aligned}$$

## 5 Jordan Canonical Form

Although every square matrix is not diagonalisable, it is possible to show that every matrix  $A$  is similar to a Jordan Form matrix  $J$ , i.e.

$$A = P^{-1} J P$$

where  $J$  is a block diagonal matrix

$$J = \text{diag}\{J_1, J_2, \dots, J_s\} = \begin{pmatrix} J_1 & 0 & \cdots & 0 \\ 0 & J_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & J_s \end{pmatrix}$$

with each block  $J_i$  being of size  $m_i \times m_i$  with  $\sum m_i = m$  and of form

$$J_i = \begin{pmatrix} \lambda & 1 & 0 & \cdots & 0 & 0 \\ 0 & \lambda & 1 & \cdots & 0 & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & 0 & \cdots & \lambda & 1 \\ 0 & 0 & 0 & \cdots & 0 & \lambda \end{pmatrix}$$

where  $\lambda$  belongs to the spectrum of  $A$ . (The same eigenvalue may appear in more than one block of  $J$ ). The Jordan Form of  $A$  is unique up to the ordering of the blocks. If  $A$  is diagonalisable, then its Jordan Form coincides with its diagonalised form.

It is straightforward to establish that

$$A^n = P^{-1} J^n P \tag{9}$$

where  $J^n = \text{diag}\{J_1^n, J_2^n, \dots, J_s^n\}$  and

$$J_i^n = \begin{pmatrix} \lambda^n & c_n(1)\lambda^{n-1} & c_n(2)\lambda^{n-2} & \cdots & c_n(m_i-1)\lambda^{n-m_i+1} \\ 0 & \lambda^n & c_n(1)\lambda^{n-1} & \cdots & c_n(m_i-2)\lambda^{n-m_i+2} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \cdots & \lambda^n \end{pmatrix} \tag{10}$$

where  $c_n(j) = \binom{n}{j}$ . Example:

$$\hat{A} = \begin{pmatrix} 0 & 1 \\ -1 & -2 \end{pmatrix}$$

has eigenvalue  $\lambda = -1$  (multiplicity 2) and associated eigenvector  $\begin{pmatrix} 1 \\ -1 \end{pmatrix}$ . Hence it is not diagonalisable since there are not two linear independent eigenvectors. However ( $\hat{A} = P^{-1} J P$ )

$$\begin{pmatrix} 0 & 1 \\ -1 & -2 \end{pmatrix} = \begin{pmatrix} 1 & -1 \\ -1 & 2 \end{pmatrix} \begin{pmatrix} -1 & 1 \\ 0 & -1 \end{pmatrix} \begin{pmatrix} 2 & 1 \\ 1 & 1 \end{pmatrix}$$

and ( $\hat{A}^n = P^{-1} J^n P$ )

$$\begin{aligned} \begin{pmatrix} 0 & 1 \\ -1 & -2 \end{pmatrix}^n &= \begin{pmatrix} 1 & -1 \\ -1 & 2 \end{pmatrix} \begin{pmatrix} (-1)^n & n(-1)^{n-1} \\ 0 & (-1)^n \end{pmatrix} \begin{pmatrix} 2 & 1 \\ 1 & 1 \end{pmatrix} \\ &= \begin{pmatrix} (-1)^n + n(-1)^{n-1} & n(-1)^{n-1} \\ -n(-1)^{n-1} & (-1)^n - n(-1)^{n-1} \end{pmatrix} \end{aligned}$$

## 6 The Discrete Putzer Algorithm

The Cayley-Hamilton Theorem states that every square matrix obeys its own characteristic equation, i.e.

$$A^m + a_{m-1}A^{m-1} + \dots + a_1A + a_0I = \mathbf{0} = (A - \lambda_1 I)(A - \lambda_2 I) \dots (A - \lambda_m I)$$

Example:  $\hat{A}$  in the previous example has characteristic polynomial  $\lambda^2 + 2\lambda + 1$ . Hence

$$\hat{A}^2 + 2\hat{A} + I = \begin{pmatrix} -1 & -2 \\ 2 & 3 \end{pmatrix} + \begin{pmatrix} 0 & 2 \\ -2 & -4 \end{pmatrix} + \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} = \mathbf{0}$$

This relationship is the basis of a number of methods for calculating  $A^n$ , none of which require knowledge of the Jordan Form of  $A$ . One of these is the Discrete Putzer Algorithm. Define the sequence of matrices  $M_i$  by

$$M_0 = I \quad M_{i+1} = (A - \lambda_{i+1}I)M_i, \quad i = 0, 1, \dots, m-1 \quad (11)$$

and the scalar sequences  $\{s_i(n)\}$  by

$$s_1(n) = \lambda_1^n \quad (12)$$

$$s_i(n) = \sum_{j=0}^{n-1} \lambda_i^{n-1-j} s_{i-1}(j), \quad i = 2, 3, \dots, m \quad (13)$$

Then it may be shown that

$$A^n = \sum_{i=1}^m s_i(n) M_{i-1} \quad (14)$$

Example: Again for  $\hat{A}$ ,  $\lambda_1 = -1$ ,  $\lambda_2 = -1$  and so

$$\begin{aligned} M_0 &= I = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \\ M_1 &= (\hat{A} - \lambda_1 I)I = \begin{pmatrix} 1 & 1 \\ -1 & -1 \end{pmatrix} \\ s_1(n) &= \lambda_1^n = (-1)^n \\ s_2(n) &= \sum_{j=0}^{n-1} \lambda_2^{n-1-j} s_1(j) = \sum_{j=0}^{n-1} (-1)^{n-1-j} (-1)^j = n(-1)^{n-1} \end{aligned}$$

Hence

$$\begin{aligned} \hat{A}^n &= \sum_{i=1}^2 s_i(n) M_{i-1} = s_1(n) M_0 + s_2(n) M_1 \\ &= (-1)^n \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} + n(-1)^{n-1} \begin{pmatrix} 1 & 1 \\ -1 & -1 \end{pmatrix} \\ &= \begin{pmatrix} (-1)^n + n(-1)^{n-1} & n(-1)^{n-1} \\ -n(-1)^{n-1} & (-1)^n - n(-1)^{n-1} \end{pmatrix} \end{aligned}$$