

## Section 5.2 The Characteristic Equation

Review:

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

Find eigenvectors  $\mathbf{x}$  by solving  $(A - \lambda I)\mathbf{x} = \mathbf{0}$ .

How do we find the eigenvalues  $\lambda$ ?

$\mathbf{x}$  must be nonzero

⇓

$(A - \lambda I)\mathbf{x} = \mathbf{0}$  must have nontrivial solutions

⇓

$(A - \lambda I)$  is not invertible

⇓

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

(called the *characteristic equation*)

Solve  $\det(A - \lambda I) = 0$  for  $\lambda$  to find the eigenvalues.

*Characteristic polynomial:*  $\det(A - \lambda I)$

*Characteristic equation:*  $\det(A - \lambda I) = 0$

**EXAMPLE:** Find the eigenvalues of  $A = \begin{bmatrix} 0 & 1 \\ -6 & 5 \end{bmatrix}$ .

*Solution:* Since

$$A - \lambda I = \begin{bmatrix} 0 & 1 \\ -6 & 5 \end{bmatrix} - \begin{bmatrix} \lambda & 0 \\ 0 & \lambda \end{bmatrix} = \begin{bmatrix} -\lambda & 1 \\ -6 & 5 - \lambda \end{bmatrix},$$

the equation  $\det(A - \lambda I) = 0$  becomes

$$-\lambda(5 - \lambda) + 6 = 0$$

$$\lambda^2 - 5\lambda + 6 = 0$$

Factor:

$$(\lambda - 2)(\lambda - 3) = 0.$$

So the eigenvalues are 2 and 3.

For a  $3 \times 3$  matrix or larger, recall that a determinant can be computed by cofactor expansion.

**EXAMPLE:** Find the eigenvalues of  $A = \begin{bmatrix} 1 & 2 & 1 \\ 0 & -5 & 0 \\ 1 & 8 & 1 \end{bmatrix}$ .

*Solution:*

$$A - \lambda I = \begin{bmatrix} 1 - \underline{\quad} & 2 & 1 \\ 0 & -5 - \underline{\quad} & 0 \\ 1 & 8 & 1 - \underline{\quad} \end{bmatrix}$$

$$\det(A - \lambda I) = \begin{vmatrix} 1 - \lambda & 2 & 1 \\ 0 & -5 - \lambda & 0 \\ 1 & 8 & 1 - \lambda \end{vmatrix} = (-5 - \lambda) \begin{vmatrix} 1 - \lambda & 1 \\ 1 & 1 - \lambda \end{vmatrix}$$

$$= (-5 - \lambda)[(1 - \lambda)^2 - 1] = (-5 - \lambda)[1 - 2\lambda + \lambda^2 - 1]$$

$$= (-5 - \lambda)[-2\lambda + \lambda^2] = -(5 + \lambda)\lambda[-2 + \lambda] = 0$$

$$\Rightarrow \lambda = -5, 0, 2$$

**THEOREM** (The Invertible Matrix Theorem - continued)

Let  $A$  be an  $n \times n$  matrix. Then  $A$  is invertible if and only if:

- s. The number 0 is not an eigenvalue of  $A$ .
- t.  $\det A \neq 0$

Recall that if  $B$  is obtained from  $A$  by a sequence of row replacements or interchanges, but without scaling, then  $\det A = (-1)^r \det B$ , where  $r$  is the number of row interchanges.

Suppose the echelon form  $U$  is obtained from  $A$  by a sequence of row replacements or interchanges, but without scaling.

$$A \sim U = \begin{bmatrix} u_{11} & u_{12} & u_{13} & \cdots & u_{1n} \\ 0 & u_{22} & u_{23} & \cdots & u_{2n} \\ 0 & 0 & u_{33} & \cdots & u_{3n} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & 0 & u_{nn} \end{bmatrix}$$

The **determinant** of  $A$ , written  $\det A$ , is defined as follows:

$$\det A = \begin{cases} (-1)^r \cdot \left( \begin{array}{l} \text{product of} \\ \text{pivots in } U \end{array} \right), & \text{when } A \text{ is invertible} \\ 0, & \text{when } A \text{ is not invertible} \end{cases}$$

( $r$  is the number of row interchanges)

**EXAMPLE:** Find the eigenvalues of  $A = \begin{bmatrix} 3 & 2 & 3 \\ 0 & 6 & 10 \\ 0 & 0 & 2 \end{bmatrix}$ .

*Solution:*

$$\det(A - \lambda I) = \det \begin{bmatrix} 3 - \lambda & 2 & 3 \\ 0 & 6 - \lambda & 10 \\ 0 & 0 & 2 - \lambda \end{bmatrix}$$

Characteristic equation:

$$(\quad)(\quad)(\quad) = 0.$$

**eigenvalues:** \_\_\_\_\_, \_\_\_\_\_, \_\_\_\_\_

The **(algebraic) multiplicity** of an eigenvalue is its multiplicity as a root of the characteristic equation.

**EXAMPLE:** Find the characteristic polynomial of

$$A = \begin{bmatrix} 2 & 0 & 0 & 0 \\ 5 & 3 & 0 & 0 \\ 9 & 1 & 3 & 0 \\ 1 & 2 & 5 & -1 \end{bmatrix}$$

and then find all the eigenvalues and the algebraic multiplicity of each eigenvalue.

Solution:

$$\det(A - \lambda I) = \begin{vmatrix} 2 - \lambda & 0 & 0 & 0 \\ 5 & 3 - \lambda & 0 & 0 \\ 9 & 1 & 3 - \lambda & 0 \\ 1 & 2 & 5 & -1 - \lambda \end{vmatrix}$$

$$= (2 - \lambda)(3 - \lambda)(3 - \lambda)(-1 - \lambda) = 0$$

**eigenvalues:** \_\_\_\_\_, \_\_\_\_\_, \_\_\_\_\_

## Similarity

Numerical methods for finding approximating eigenvalues are based upon Theorem 4 to be described shortly.

For  $n \times n$  matrices  $A$  and  $B$ , we say the  $A$  is **similar** to  $B$  if there is an invertible matrix  $P$  such that

$$P^{-1}AP = B \quad \text{or equivalently,} \quad A = PBP^{-1}.$$

**Theorem 4:** If  $n \times n$  matrices  $A$  and  $B$  are similar, then they have the same characteristic polynomial and hence the same eigenvalues (with the same multiplicities).

**Proof:** If  $B = P^{-1}AP$ , then

$$\begin{aligned} \det(B - \lambda I) &= \det[P^{-1}AP - P^{-1}\lambda IP] = \det[P^{-1}(A - \lambda I)P] \\ &= \det P^{-1} \cdot \det(A - \lambda I) \cdot \det P = \det(A - \lambda I). \end{aligned}$$

## Application to Markov Chains

**EXAMPLE** Consider the migration matrix  $M = \begin{bmatrix} .95 & .90 \\ .05 & .10 \end{bmatrix}$  and define  $\mathbf{x}_{k+1} = M\mathbf{x}_k$ . It can be shown that

$$\mathbf{x}_0, \mathbf{x}_1, \mathbf{x}_2, \dots$$

converges to a steady state vector  $\mathbf{x} = \begin{bmatrix} \frac{1}{2} \\ \frac{1}{2} \end{bmatrix}$ . Why?

The answer lies in examining the corresponding eigenvectors.

First we find the eigenvalues:

$$\det(M - \lambda I) = \det\left(\begin{bmatrix} .95 - \lambda & .90 \\ .05 & .10 - \lambda \end{bmatrix}\right) = \lambda^2 - 1.05\lambda + 0.05$$

So solve

$$\lambda^2 - 1.05\lambda + 0.05 = 0$$

By factoring

$$\lambda = 0.05, \lambda = 1$$

It can be shown that the eigenspace corresponding to  $\lambda = 1$  is  $\text{span}\{\mathbf{v}_1\}$  where  $\mathbf{v}_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$  and the eigenspace corresponding to  $\lambda = 0.05$  is  $\text{span}\{\mathbf{v}_2\}$  where  $\mathbf{v}_2 = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$ .

Note that

$$M\mathbf{v}_1 = \mathbf{v}_1,$$

and so  $\begin{bmatrix} \frac{1}{2} \\ \frac{1}{2} \end{bmatrix}$  is our steady state vector.

Then for a given vector  $\mathbf{x}_0$ ,

$$\mathbf{x}_0 = c_1\mathbf{v}_1 + c_2\mathbf{v}_2$$

$$\mathbf{x}_1 = M\mathbf{x}_0 = M(c_1\mathbf{v}_1 + c_2\mathbf{v}_2) = c_1M\mathbf{v}_1 + c_2M\mathbf{v}_2 = c_1\mathbf{v}_1 + c_2(0.05)\mathbf{v}_2$$

$$\mathbf{x}_2 = M\mathbf{x}_1 = M(c_1\mathbf{v}_1 + c_2(0.05)\mathbf{v}_2) = c_1M\mathbf{v}_1 + c_2(0.05)M\mathbf{v}_2 = c_1\mathbf{v}_1 + c_2(0.05)^2\mathbf{v}_2$$

and in general

$$\mathbf{x}_k = c_1\mathbf{v}_1 + c_2(0.05)^k\mathbf{v}_2$$

$$\text{and so } \lim_{k \rightarrow \infty} \mathbf{x}_k = \lim_{k \rightarrow \infty} (c_1\mathbf{v}_1 + c_2(0.05)^k\mathbf{v}_2) = c_1\mathbf{v}_1$$

and this is the steady state when  $c_1 = \frac{1}{2}$ .