

Lesson 6: Diagonalization

Algebra

Computer Science Engineering Degree

May 4, 2016

1 Introductory example

2 Diagonalization of matrices

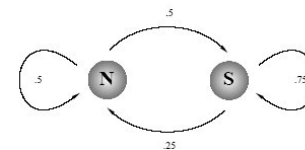
Index

1 Introductory example

2 Diagonalization of matrices

Problem

Suppose that the population moves from two geographical regions, say the North and South, as follows: each year 50% of the North population migrates to the south, while 25% of the South population migrates to the North.



Problema

If this migration pattern continues, will the North population decrease until all population will be concentrated in the South, or is there a stabilization tendency?

Statement or the problem using matrices

Let us denote n_0 and s_0 the proportions of initial populations in the North and the South, respectively. In the same way, let us denote n_k and s_k the proportions at the end of the k th year. Observe that $n_k + s_k = 1$.

The migration pattern shows that

$$\left. \begin{aligned} n_{k+1} &= 0,5 n_k + 0,25 s_k \\ s_{k+1} &= 0,5 n_k + 0,75 s_k \end{aligned} \right\}$$

Using matrices:

$$\begin{bmatrix} n_{k+1} \\ s_{k+1} \end{bmatrix} = \begin{bmatrix} 0,5 & 0,25 \\ 0,5 & 0,75 \end{bmatrix} \begin{bmatrix} n_k \\ s_k \end{bmatrix}$$

Statement of the problem using matrices

Let

$$\vec{x}_k = \begin{bmatrix} n_k \\ s_k \end{bmatrix} \quad \text{y} \quad A = \begin{bmatrix} 0,5 & 0,25 \\ 0,5 & 0,75 \end{bmatrix}$$

Then: $\vec{x}_{k+1} = A\vec{x}_k$ para todo $k = 0, 1, 2, \dots$, es decir,

$$\vec{x}_1 = A\vec{x}_0, \quad \vec{x}_2 = A\vec{x}_1 = A^2\vec{x}_0, \quad \vec{x}_3 = A\vec{x}_2 = A^3\vec{x}_0, \quad \dots$$

That is:

$$\vec{x}_k = A^k \vec{x}_0 \text{ for all } k \geq 1$$

Therefore, the successive powers of A determine the process.

Task

To give an expression of A^k for all k and, using it, to study the behaviour of \vec{x}_k when k is big.

Idea

The theory we shall see in this lesson will allow us to obtain an **invertible** matrix P and a **diagonal** matrix D such that $P^{-1}AP = D$. In fact, we can take:

$$P = \begin{bmatrix} 1 & -1 \\ 2 & 1 \end{bmatrix} \quad \text{y} \quad D = \begin{bmatrix} 1 & 0 \\ 0 & 1/4 \end{bmatrix}$$

Then $A = PDP^{-1}$ and, therefore:

$$A^k = \underbrace{PDP^{-1}} \underbrace{PDP^{-1}} \dots \underbrace{PDP^{-1}} =$$

$$P \cdot D^k \cdot P^{-1} = P \begin{bmatrix} 1^k & 0 \\ 0 & (1/4)^k \end{bmatrix} P^{-1} = P \begin{bmatrix} 1 & 0 \\ 0 & 1/4^k \end{bmatrix} P^{-1}$$

Solution

Since

$$\vec{x}_k = A^k \vec{x}_0 = P \begin{bmatrix} 1 & 0 \\ 0 & 1/4^k \end{bmatrix} P^{-1} \begin{bmatrix} n_0 \\ s_0 \end{bmatrix}$$

taking limits when $k \rightarrow \infty$ we have:

$$\begin{aligned} \lim_{k \rightarrow \infty} \vec{x}_k &= P \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} P^{-1} \begin{bmatrix} n_0 \\ s_0 \end{bmatrix} = \\ &= \begin{bmatrix} 1/3 & 1/3 \\ 2/3 & 2/3 \end{bmatrix} \begin{bmatrix} n_0 \\ s_0 \end{bmatrix} = \begin{bmatrix} (1/3)n_0 + (1/3)s_0 \\ (2/3)n_0 + (2/3)s_0 \end{bmatrix} \\ &= \begin{bmatrix} (1/3)(n_0 + s_0) \\ (2/3)(n_0 + s_0) \end{bmatrix} = \begin{bmatrix} 1/3 \\ 2/3 \end{bmatrix} \end{aligned}$$

since $n_0 + s_0 = 1$. Then, the long-term tendency is a stabilization of 1/3 of the population in the North and 2/3 in the South, **independently of the initial distribution** (n_0, s_0) .

1 Introductory example

2 Diagonalization of matrices

Similarity of matrices and diagonalizable matrices

Definition

Two square matrices A and B are *similar* if there exists an invertible matrix P such that $A = PBP^{-1}$.

This is an equivalence relation in the set of square matrices.

Definition

A square matrix A is *diagonalizable* if it is similar to a diagonal matrix.

Problem to analyze: Given a square matrix A is it diagonalizable? In the affirmative case, how can we compute a diagonal matrix D and an invertible matrix P such that $A = PDP^{-1}$?

Eigenvalues and eigenvectors

Definition

Let A be a square matrix.

- We shall say that a scalar λ is an **eigenvalue** of A if there exists a **non-zero** vector \vec{x} such that $A\vec{x} = \lambda\vec{x}$. In this case, each vector $\vec{x} \neq \vec{0}$ satisfying this condition is called **eigenvector** associated with λ .
- Given an eigenvalue λ of A , we call **eigenspace** associated with λ to the vector subspace V_λ whose elements are $\vec{0}$ and all the eigenvectors associated to λ , that is, $V_\lambda := \{\vec{x} \in \mathbb{R}^n \mid A\vec{x} = \lambda\vec{x}\} = \text{Ker}(A - \lambda I)$.

Computation of eigenvalues

If A is a square matrix:

λ is an eigenvalue of $A \iff$

There exists $\vec{x} \neq \vec{0}$ such that $A\vec{x} = \lambda\vec{x} \iff$

There exists $\vec{x} \neq \vec{0}$ such that $(A - \lambda I)\vec{x} = \vec{0} \iff$

The homogeneous system $(A - \lambda I)\vec{x} = \vec{0}$ has non-zero solutions \iff

The homogeneous system $(A - \lambda I)\vec{x} = \vec{0}$ has infinitely many solutions \iff

The matrix $A - \lambda I$ is not invertible \iff

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

Computation of eigenvalues

Definition

Given a square matrix A , its *characteristic polynomial* is

$$p_A(\lambda) := \det(A - \lambda I) = \begin{vmatrix} a_{11} - \lambda & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} - \lambda & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} - \lambda \end{vmatrix}$$

From the previous slide:

Theorem

The eigenvalues of A are the **roots of the characteristic polynomial**.

Some immediate properties

- (1) $p_A(\lambda)$ is a polynomial of degree n , the order of A .
- (2) There exist, at most, n different eigenvalues.
- (3) Two similar matrices have the same characteristic polynomial.

Exercise 1: Prove these properties.

EXAMPLE: Consider the matrix $A = \begin{bmatrix} 2 & 2 \\ 1 & 1 \end{bmatrix}$. Its characteristic polynomial is:

$$p_A(\lambda) = \begin{vmatrix} 2 - \lambda & 2 \\ 1 & 1 - \lambda \end{vmatrix} = (2 - \lambda)(1 - \lambda) - 2 = \lambda^2 - 3\lambda.$$

Then A has 2 different eigenvalues $\lambda_1 = 0$ and $\lambda_2 = 3$.

Examples

EXAMPLE: Consider the matrix $A = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}$. Its characteristic polynomial is:

$$p_A(\lambda) = \begin{vmatrix} -\lambda & 1 \\ -1 & -\lambda \end{vmatrix} = \lambda^2 + 1.$$

Since it has no real roots, A **has not eigenvalues**.

EXAMPLE: If $A = \begin{bmatrix} -2 & 4 & 5 \\ -3 & 5 & 5 \\ 0 & 0 & 1 \end{bmatrix}$ then

$$p_A(\lambda) = \begin{vmatrix} -2 - \lambda & 4 & 5 \\ -3 & 5 - \lambda & 5 \\ 0 & 0 & 1 - \lambda \end{vmatrix} = (-2 - \lambda)(5 - \lambda)(1 - \lambda) + 12(1 - \lambda) =$$

$$-\lambda^3 + 4\lambda^2 - 5\lambda + 2 = -(\lambda - 1)^2(\lambda - 2)$$

Therefore A has 2 different eigenvalues: $\lambda_1 = 1$ and $\lambda_2 = 2$.

Algebraic and geometric multiplicity

Definition

Let A a square matrix of order n and λ_i an eigenvalue.

- The *algebraic multiplicity* of λ_i is its multiplicity as a root of the characteristic polynomial of A , that is, the greatest exponent α_i for which the factor $(\lambda - \lambda_i)^{\alpha_i}$ appears in the decomposition of $p_A(\lambda)$.
- The *geometric multiplicity* of λ_i is the dimension d_i of its associated eigenspace, that is, $d_i = \dim V_{\lambda_i}$.

Since $V_{\lambda_i} = \text{Ker}(A - \lambda_i I)$, by a formula of Lesson 4 we have:

Proposición

$$d_i = n - \text{rank}(A - \lambda_i I),$$

Algebraic and geometric multiplicities

EXAMPLE: Consider the matrix $A = \begin{bmatrix} 2 & 2 & 3 \\ 0 & 2 & 2 \\ 0 & 0 & 2 \end{bmatrix}$.

The characteristic polynomial is:

$$p_A(\lambda) = \begin{vmatrix} 2-\lambda & 2 & 3 \\ 0 & 2-\lambda & 2 \\ 0 & 0 & 2-\lambda \end{vmatrix} = (2-\lambda)^3.$$

A has a unique eigenvalue $\lambda_1 = 2$ whose algebraic multiplicity is $\alpha_1 = 3$.
The geometric multiplicity is:

$$d_1 = 3 - \text{rank}(A - 2I) = 3 - \text{rank} \begin{bmatrix} 0 & 2 & 3 \\ 0 & 0 & 2 \\ 0 & 0 & 0 \end{bmatrix} = 3 - 2 = 1.$$

Properties of the eigenspaces

Let A a square matrix of order n and let $\lambda_1, \lambda_2, \dots, \lambda_r$ be its eigenvalues.

- (a) $d_i = \dim V_{\lambda_i} \geq 1$ for all i .
- (b) $V_{\lambda_i} \cap (V_{\lambda_i} + \dots + V_{\lambda_{i-1}} + V_{\lambda_{i+1}} + \dots + V_{\lambda_r}) = \{\vec{0}\}$ for all i . In particular $V_{\lambda_i} \cap V_{\lambda_j} = \{\vec{0}\}$ si $i \neq j$.
- (c) Eigenvectors associated to different eigenvalues are linearly independent.
- (d) If \mathcal{B}_i is a basis of V_{λ_i} (for all $i = 1, \dots, r$) then $\mathcal{B}_1 \cup \dots \cup \mathcal{B}_r$ is a **basis** of $V_{\lambda_1} + \dots + V_{\lambda_r}$.
- (f) The following inequality is satisfied (**very useful!!!**): $1 \leq d_i \leq \alpha_i$ for all $i = 1, 2, \dots, r$.

Diagonalization criteria

Theorem (first criterion of diagonalization)

A square matrix A of order n is diagonalizable if and only if $V_{\lambda_1} + V_{\lambda_2} + \dots + V_{\lambda_r} = \mathbb{R}^n$, where $\lambda_1, \lambda_2, \dots, \lambda_r$ are all the distinct eigenvalues of A (using (d) this is equivalent to say that **there exists a basis of \mathbb{R}^n whose elements are eigenvectors of A**).

Theorem (second criterion of diagonalization) (**extremely important!!!**)

If A and $\lambda_1, \dots, \lambda_r$ are as in the previous theorem, A is diagonalizable if and only if these two properties are satisfied:

- (1) $\alpha_1 + \alpha_2 + \dots + \alpha_r = n$ (that is, all the roots of $p_A(\lambda)$ are real),
- (2) $d_i = \alpha_i$ for all $i = 1, 2, \dots, r$.

As a consequence of this criterion and Property (f) of the previous slide:

Corollary (**very useful!!!**)

If a square matrix of order n has n distinct (real) eigenvalues then it is diagonalizable.

Diagonalization algorithm

Steps to determine if a square matrix A of order n is diagonalizable:

- (1) Compute the characteristic polynomial $p_A(\lambda)$.
- (2) Decompose into factors the characteristic polynomial, obtaining the eigenvalues $\lambda_1, \dots, \lambda_r$ and their algebraic multiplicities $\alpha_1, \dots, \alpha_r$. **If $\alpha_1 + \dots + \alpha_r < n$ therefore A is not diagonalizable** (by the second criterion). Otherwise go to the next step
- (3) Compute the geometric multiplicities: $d_i = n - \text{rank}(A - \lambda_i I)$. **If, for some i , $d_i \neq \alpha_i$ then the matrix is not diagonalizable** (by the second criterion). Otherwise A is diagonalizable.

Diagonalization algorithm

Steps to compute the matrices P and D such that D is diagonal and $P^{-1}AP = D$:

- (4) The diagonal elements of D are $\overbrace{\lambda_1, \lambda_1, \dots, \lambda_1}^{\alpha_1}, \overbrace{\lambda_2, \lambda_2, \dots, \lambda_2}^{\alpha_2}, \dots, \overbrace{\lambda_r, \lambda_r, \dots, \lambda_r}^{\alpha_r}$ (each eigenvalue appears repeated according with its algebraic multiplicity).
- (5) Obtain bases \mathcal{B}_i of the eigenspaces V_{λ_i} .
- (6) The matrix P is the one whose columns are the components of the vectors of the basis of \mathbb{R}^n given by $\mathcal{B}_1 \cup \mathcal{B}_2 \cup \dots \cup \mathcal{B}_r$ (ordered accordingly with the order of the eigenvalues in the diagonal of D).

Example

The next step consists of computing the geometric multiplicities:

$$d_1 = 3 - \text{rank}(A - 2I) = 3 - \text{rank} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} = 3 - 1 = 2.$$

For d_2 , applying that $1 \leq d_2 \leq \alpha_2 = 1$, we obtain directly $d_2 = 1$. We have, then, that $d_1 = \alpha_1$ and $d_2 = \alpha_2$. Applying the second criterion, the matrix A is diagonalizable and its diagonal form is

$$D = \begin{bmatrix} 2 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 5 \end{bmatrix}$$

Example

We shall study if the matrix $A = \begin{bmatrix} 3 & 1 & 1 \\ 1 & 3 & 1 \\ 1 & 1 & 3 \end{bmatrix}$ is diagonalizable and we shall compute, in the affirmative case, the matrices D and P . We begin computing the characteristic polynomial of A :

$$p_A(\lambda) = \begin{vmatrix} 3 - \lambda & 1 & 1 \\ 1 & 3 - \lambda & 1 \\ 1 & 1 & 3 - \lambda \end{vmatrix} = -\lambda^3 + 9\lambda^2 - 24\lambda + 20.$$

We decompose $p_A(\lambda)$ using Ruffini:

$$p_A(\lambda) = (2 - \lambda)^2(5 - \lambda).$$

Then the eigenvalues of A and their algebraic multiplicities are:

$$\begin{aligned} \lambda_1 &= 2, & \alpha_1 &= 2 \\ \lambda_2 &= 5, & \alpha_2 &= 1 \end{aligned}$$

Ejemplo

To compute the matrix P we need the bases of the eigenspaces V_{λ_1} and V_{λ_2} . We compute a basis of V_{λ_1} :

Since $A - 2I = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$ it holds that

$$(x, y, z) \in V_{\lambda_1} \Leftrightarrow \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \Leftrightarrow \{ x + y + z = 0 \}$$

Therefore, a basis of V_{λ_1} is $\{(-1, 1, 0), (-1, 0, 1)\}$.

Example

In the same way: $A - 5I = \begin{bmatrix} -2 & 1 & 1 \\ 1 & -2 & 1 \\ 1 & 1 & -2 \end{bmatrix}$ and therefore

$$(x, y, z) \in V_{\lambda_2} \Leftrightarrow \begin{bmatrix} 2 & 1 & 1 \\ 1 & -2 & 1 \\ 1 & 1 & -2 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \Leftrightarrow \begin{cases} -2x + y + z = 0 \\ x - 2y + z = 0 \\ x + y - 2z = 0 \end{cases}$$

Solving the system one obtains that a basis of V_{λ_2} is $\{(1, 1, 1)\}$.

As a consequence, a basis of \mathbb{R}^3 given by eigenvectors is $\{(-1, 1, 0), (-1, 0, 1), (1, 1, 1)\}$. The matrix P is

$$P = \begin{bmatrix} -1 & -1 & 1 \\ 1 & 0 & 1 \\ 0 & 1 & 1 \end{bmatrix}$$

Some interesting properties

- (1) If A and B are similar matrices then $|A| = |B|$ and, moreover, they have the same rank.
- (2) If A is a *symmetric matrix* then it is diagonalizable.
- (3) If A is triangular, its eigenvalues are the diagonal elements.
- (4) If λ is an eigenvalue of A then λ^k is an eigenvalue of A^k for all $k \in \mathbb{N}$.