| 5 | Eige | Eigenvalues and Eigenvectors | |
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| | 5.1 | Introduction | |
| | 5.2 | Diagonalization of a Matrix | |
| | 5.3 | Difference Equations and Powers A^k | |
| | 5.4 | Differential Equations and e^{At} | |
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5

Eigenvalues and Eigenvectors (特征值与特征向量)

5.1

EIGENVALUES AND EIGENVECTORS

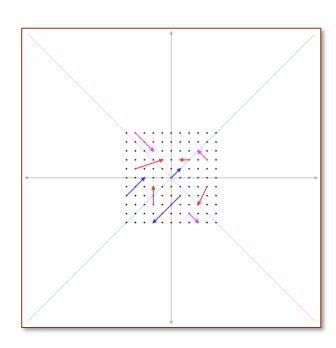
Introduction

Definition

Calculations

Properties

Application



I. Introductory example - 人口流动问题







人口流动

城市建设

家庭经济条件

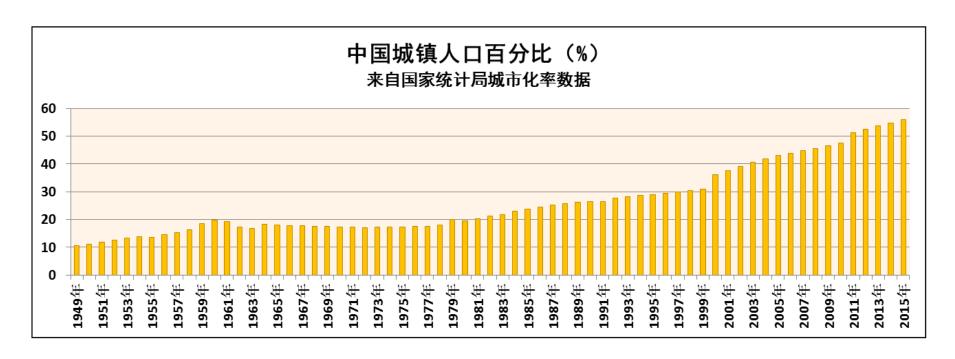
农村劳动力

留守儿童

空巢老人

春运

城市管理



发展趋势?

简化模型

● 假设:



从事农业工作的人员中每年有四分之一转为从事非农业工作, 从事非农业工作的人员中每年 有六分之一转为从事农业工作.





- ❷ 人口总数不变.
 - 🦿 预测多年之后劳动力从业情况的发展趋势.
- 分析 设最初农业人员和非农业人员的数量分别为 y_0, z_0 ,第 1 年末数量为 y_1, z_1 ,

$$\begin{cases} y_1 = \frac{3}{4} y_0 + \frac{1}{6} z_0, \\ z_1 = \frac{1}{4} y_0 + \frac{5}{6} z_0. \end{cases} \qquad \qquad \begin{bmatrix} y_1 \\ z_1 \end{bmatrix} = \begin{bmatrix} \frac{3}{4} & \frac{1}{6} \\ \frac{1}{4} & \frac{5}{6} \end{bmatrix} \begin{bmatrix} y_0 \\ z_0 \end{bmatrix}$$

分析 设最初农业人员和非农业人员的数量分别为 y_0, z_0 ,第 1 年末数量为 y_1, z_1 ,

$$\begin{bmatrix} y_1 \\ z_1 \end{bmatrix} = \begin{bmatrix} \frac{3}{4} & \frac{1}{6} \\ \frac{1}{4} & \frac{5}{6} \end{bmatrix} \begin{bmatrix} y_0 \\ z_0 \end{bmatrix}$$

分析 设最初农业人员和非农业人员的数量分别为 y_0, z_0 , 第 k 年末数量为 y_k, z_k ,

$$\begin{bmatrix} y_1 \\ z_1 \end{bmatrix} = \begin{bmatrix} \frac{3}{4} & \frac{1}{6} \\ \frac{1}{4} & \frac{5}{6} \end{bmatrix} \begin{bmatrix} y_0 \\ z_0 \end{bmatrix}$$

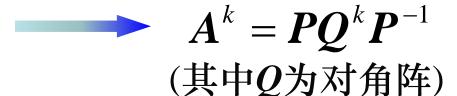
如何计算 方阵 A 的幂 A^k ?

$$\begin{bmatrix} y_k \\ z_k \end{bmatrix} = \begin{bmatrix} \frac{3}{4} & \frac{1}{6} \\ \frac{1}{4} & \frac{5}{6} \end{bmatrix} \begin{bmatrix} y_{k-1} \\ z_{k-1} \end{bmatrix} = \dots = \begin{bmatrix} \frac{3}{4} & \frac{1}{6} \\ \frac{1}{4} & \frac{5}{6} \end{bmatrix} \begin{bmatrix} y_0 \\ z_0 \end{bmatrix} = A^k \begin{bmatrix} y_0 \\ z_0 \end{bmatrix}$$

最简单的方阵
$$Q$$
: $Q = \begin{bmatrix} \lambda_1 \\ \lambda_2 \end{bmatrix}$, $Q^k = \begin{bmatrix} \lambda_1^k \\ \lambda_2^k \end{bmatrix}$.

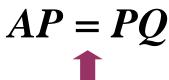
有无可能?

$$A = PQP^{-1}$$



$$A^{2} = A A = PQP^{-1}PQP^{-1} = PQ^{2}P^{-1}$$

 $A^{3} = A A A = PQP^{-1}PQP^{-1}PQP^{-1} = PQ^{3}P^{-1}$



$$\mathbf{A}[\mathbf{P}_1 \mid \mathbf{P}_2] = [\mathbf{P}_1 \mid \mathbf{P}_2] \begin{bmatrix} \lambda_1 \\ \lambda_2 \end{bmatrix} \longleftrightarrow$$

$$(其中 P_1, P_2 为2维列向量)$$

$$AP_1 = \lambda_1 P_1, AP_2 = \lambda_2 P_2$$



$$[\boldsymbol{AP}_1 \ \boldsymbol{AP}_2] = [\lambda_1 \boldsymbol{P}_1 \ \lambda_2 \boldsymbol{P}_2]$$

$$\boldsymbol{A}_{2\times2}\boldsymbol{x}_{2\times1} = \boldsymbol{\lambda}\boldsymbol{x}_{2\times1}$$

II. Eigenvalues and Eigenvectors – Definition & Calculation

Definition 1 Let A be a square matrix of degree n.

If there exist a non-zero vector \mathbf{x} and a scalar λ such that

$$A x = \lambda x$$
,

then λ is called an eigenvalue (特征值) of A, and x is called an eigenvector (特征向量), corresponding to the eigenvalue λ .



David Hilbert 1862-1943



德语 eigen (有特征的, 自身的, 个体的)



A 的特征值与特征向量有何直观含义?

$$R^n \to R^n$$

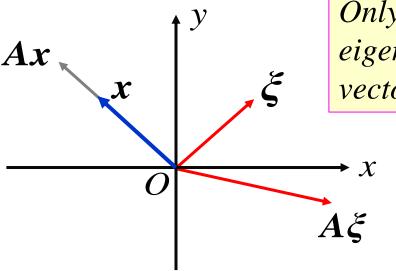
$$A: x \to Ax$$

$$R^n \to R^n$$

$$\lambda: x \to \lambda x$$

给定方阵 $A \in \mathbb{R}^{n \times n}$, 一般说来,对于 $\xi \in \mathbb{R}^n$, 向量 $A\xi$ 与 ξ <u>不在</u>同一方向上.

但也可能存在向量 x, 使得 Ax在x的方向上. Ax is a multiple of x.

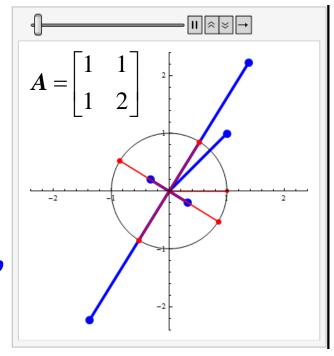


Only certain special numbers are eigenvalues, and only certain special vectors are eigenvectors.

A 的特征值与特征向量的直观含义:

特征向量:与A 左乘下的像"共线"

特 征 值:"伸缩比例"



For example,
$$A = \begin{bmatrix} 1 & 3 \\ 2 & 2 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 3 \\ 2 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 4 \\ 4 \end{bmatrix} = 4 \begin{bmatrix} 1 \\ 1 \end{bmatrix},$$

which shows that 4 is an eigenvalue of A, $(1,1)^T$ is an eigenvector for 4 (an eigenvector corresponding to 4).

$$\begin{bmatrix} 1 & 3 \\ 2 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ -\frac{2}{3} \end{bmatrix} = \begin{bmatrix} -1 \\ \frac{2}{3} \end{bmatrix} = (-1) \begin{bmatrix} 1 \\ -\frac{2}{3} \end{bmatrix},$$

which tells us that -1 is also an eigenvalue of A, and $(1,-2/3)^T$ is an eigenvector corresponding to -1.



!dea Equivalently, an eigenvalue λ and a corresponding eigenvector x satisfy:

$$(A - \lambda I) x = 0.$$

Since x is a *non-zero* vector, the matrix $A - \lambda I$ is not invertible, and thus the determinant $|A - \lambda I|$ equals 0.

注: 特征向量 x 是非零向量, 是齐次线性方程组

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

的非零解. ↑应满足

$$|A - \lambda I| = 0.$$

(characteristic equation: 特征方程)

We note that the determinant $|A - \lambda I|$ is a polynomial of degree n in λ , called the **characteristic polynomial** of A.

Definition 2 Let $A = [a_{ij}]_{n \times n}$ be a *square matrix* of degree n. Then

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

is called the characteristic polynomial (特征多项式) of A.

n 阶矩阵A的特征多项式是 λ 的 n 次多项式.

Lemma 1 A scalar λ is an eigenvalue of A if and only if λ is a root of the characteristic polynomial.

n 阶矩阵 A 的特征多项式是 λ 的 n 次多项式.



Gauss 1777-1855

代数基本定理(Fundamental theorem of algebra) 在复数范围内每个n 次复系数方程恰有n 个根.

注释: n 阶方阵 A 在复数范围内有 n 个特征值.

n 阶矩阵 A 的特征多项式在复数域上的 n 个根都是矩阵

A 的特征值, 其 k 重根叫做 k 重特征值.

当 $n \ge 5$ 时,特征多项式没有一般的求根公式.

(Galois and Abel proved that there can be no algebraic formula for the roots of a fifth-degree polynomial).

即使是三阶矩阵,一般也难以求根.

解决方法:"计算方法(数值分析)"



Galois 1811-1832



Abel 1802-1829

Example 1 Let
$$\mathbf{A} = \begin{bmatrix} 4 & -5 \\ 2 & -3 \end{bmatrix}$$
. Then
$$|\mathbf{A} - \lambda \mathbf{I}| = \begin{vmatrix} 4 - \lambda & -5 \\ 2 & -3 - \lambda \end{vmatrix} = \lambda^2 - \lambda - 2.$$

The roots are $\lambda_1 = -1$ and $\lambda_2 = 2$. Eigenvectors of **A** can be obtained as follows.

For $\lambda_1 = -1$, we solve the system of linear equations $(A - \lambda_1 I)x = (A + I)x = 0$, i.e.,

$$\begin{bmatrix} 5 & -5 \\ 2 & -2 \end{bmatrix} \mathbf{x} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}.$$

It follows that $\mathbf{x}_1 = k_1 (1,1)^{\mathrm{T}}$ ($k_1 \in \mathbf{R}, k_1 \neq 0$) are eigenvectors corresponding to the eigenvalue $\lambda_1 = -1$.

For $\lambda_2 = 2$, we solve the system of linear equations $(A - \lambda_2 I)x = (A - 2I)x = 0$, i.e.,

$$\begin{bmatrix} 2 & -5 \\ 2 & -5 \end{bmatrix} x = \begin{bmatrix} 0 \\ 0 \end{bmatrix}.$$

It follows that $x_2 = k_2(5,2)^{\mathrm{T}}$ ($k_2 \in \mathbb{R}, k_2 \neq 0$) are eigenvectors corresponding to the eigenvalue $\lambda_2 = 2$.

This example illustrates a method for finding eigenvalues and eigenvectors of a given matrix, which is important in the area of matrix theory and many applications.

A **Process** for finding eigenvalues and eigenvectors of a matrix A:

1. Compute the determinant of $A - \lambda I$.

With λ subtracted along the diagonal, this determinant is a polynomial of degree n. It starts with $(-\lambda)^n$.

2. Find the roots of this polynomial.

The n roots are the eigenvalues of A.

3. For each eigenvalue solve the equation $(A - \lambda I) x = 0$.

Since the determinant is zero, there are solutions other than x = 0. Those are the eigenvectors.

特征值与特征向量的求解 3%



第一步 计算A 的特征多项式;

第二步 求出特征多项式的全部根,即得A的全部特征值;

第三步 将每一个特征值代入相应的线性方程组进行求解,即得该特征值的特征向量.

$$A \longrightarrow |A - \lambda I| = 0 \longrightarrow (A - \lambda_i I) x = 0$$

求特征值 λ_i

求特征向量

Remark: 另一种等价求法

$$A \longrightarrow |\lambda I - A| = 0 \longrightarrow (\lambda_i I - A)x = 0$$
 求特征值 λ_i

Example 2 Everything is clear when **A** is a **diagonal matrix**:

$$A = \begin{bmatrix} 3 & 0 \\ 0 & 2 \end{bmatrix}$$
 has $\lambda_1 = 3$ with $x_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$, $\lambda_2 = 2$ with $x_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$.

On each eigenvector **A** acts like: $Ax_1 = 3x_1$ and $Ax_2 = 2x_2$.

Other vectors like $\mathbf{x} = \begin{bmatrix} 1 \\ 5 \end{bmatrix}$ are mixtures $\mathbf{x}_1 + 5\mathbf{x}_2$ of the two eigenvectors,

and **A** acts like:
$$A(x_1 + 5x_2) = 3x_1 + 10x_2 = \begin{bmatrix} 3 \\ 10 \end{bmatrix}$$
.

This is Ax for a typical vector x—not an eigenvector. But the action of A is determined by its eigenvectors and eigenvalues.

Remark: 1. The eigenvalues are on the main diagonal when *A* is *diagonal*.

$$|\mathbf{A} - \lambda \mathbf{I}| = (a_{11} - \lambda)(a_{22} - \lambda) \cdots (a_{nn} - \lambda).$$

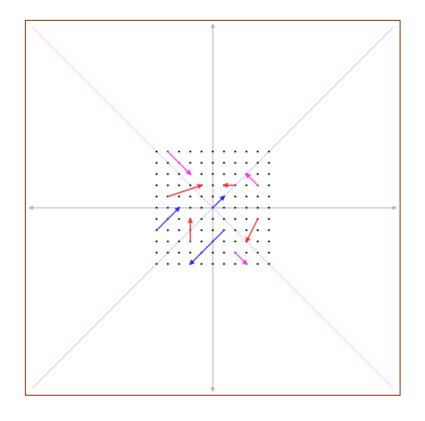
2. The eigenvalues are on the main diagonal when B is triangular. (n 阶上(下)三角形矩阵<math>B的特征值也是它的n个主对角元 b_{11} , b_{22} , \cdots , b_{nn} .)

Exercise

Find the eigenvalues and eigenvectors of $\mathbf{A} = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}$.

Example 3 Find the eigenvalues and eigenvectors of the matrix

$$A = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}$$



http://en.wikipedia.org/wiki/File:Eigenvectors-extended.gif

The Eigenvectors

$$k_{1} \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

$$k_{2} \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$$(k_{1} \neq 0)$$

$$(k_{2} \neq 0)$$

corresponding respectively to the Eigenvalues:

$$\lambda_1 = 1$$
 $\lambda_2 = 3$

Example 4 If a square matrix A satisfies $A^2 = A$. Show that the only possible eigenvalues of A are 0 or 1.

Proof Let λ be the eigenvalue of A, then $Ax = \lambda x$. And

$$A^2x = A(Ax) = A(\lambda x) = \lambda Ax = \lambda \lambda x = \lambda^2 x$$

So
$$(\lambda^2 - \lambda)x = 0$$
.

Since $x \neq 0$, we have $\lambda^2 - \lambda = 0$, and $\lambda = 0$ or $\lambda = 1$.

For example,

$$A_1 = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}, A_2 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, A_3 = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$$

Eigenvalues:

0

1

1, 0

Remark: The eigenvalues of a *projection matrix* are 1 or 0. (投影矩阵的特征值为0或1)

Remarks: Give a matrix A of degree n.

- A has exactly n eigenvalues, some of them might be repeated.
- Some eigenvalues of A may be complex numbers; some matrices may have no *real* eigenvalue, for instance, rotation matrices.

$$\mathbf{A}_{1} = \begin{bmatrix} 0 & -2 & -2 \\ 2 & -4 & -2 \\ -2 & 2 & 0 \end{bmatrix}$$

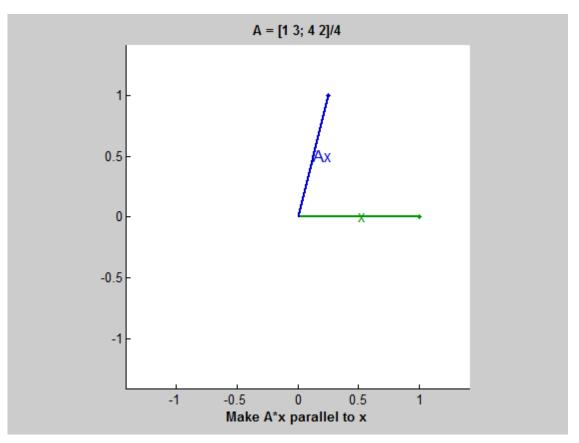
 $A_1 = \begin{bmatrix} 0 & -2 & -2 \\ 2 & -4 & -2 \end{bmatrix}$ has eigenvalues: $\lambda_1 = 0$, $\lambda_2 = -2$ (二重特征值, a root of multiplicity 2). $|A_1 - \lambda I| = -\lambda(\lambda + 2)^2$

$$|\mathbf{A}_1 - \lambda \mathbf{I}| = -\lambda(\lambda + 2)^2$$

$$A_{2} = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} \text{ has eigenvalues: } \lambda = \cos \theta \pm i \sin \theta.$$
$$|A_{2} - \lambda \mathbf{I}| = \lambda^{2} - (2\cos \theta)\lambda + 1$$

Matlab demo: eigshow

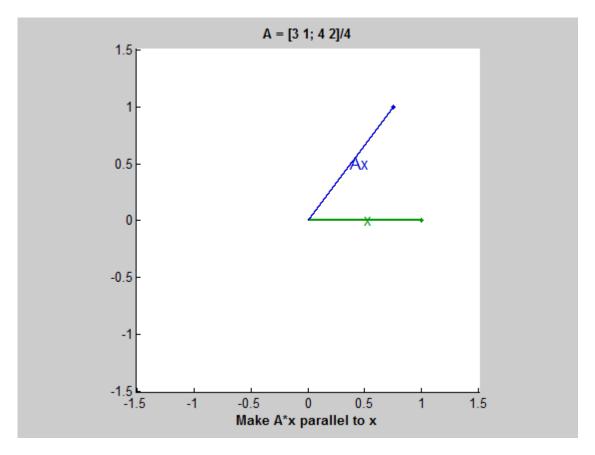
$$A = \frac{1}{4} \begin{bmatrix} 1 & 3 \\ 4 & 2 \end{bmatrix}$$
 $\lambda_1 = -0.5,$ $\lambda_2 = 1.25.$



https://blogs.mathworks.com/cleve/2013/07/08/eigshow-week-1/#f96996aa-ef86-4137-b343-07584baff36c

Matlab demo: eigshow

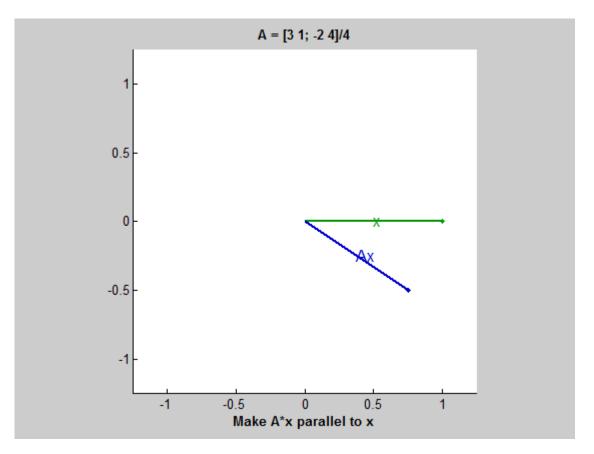
$$A = \frac{1}{4} \begin{bmatrix} 3 & 1 \\ 4 & 2 \end{bmatrix}$$
 $\lambda_1 = 1.14,$ $\lambda_2 = 0.11.$



https://blogs.mathworks.com/cleve/2013/07/08/eigshow-week-1/#f96996aa-ef86-4137-b343-07584baff36c

Matlab demo: eigshow

$$A = \frac{1}{4} \begin{bmatrix} 3 & 1 \\ -2 & 4 \end{bmatrix}$$
 $\lambda_1 = 0.8750 + 0.3307i,$ $\lambda_2 = 0.8750 - 0.3307i.$



https://blogs.mathworks.com/cleve/2013/07/08/eigshow-week-1/#f96996aa-ef86-4137-b343-07584baff36c

III. Eigenvalues and Eigenvectors – Properties

Theorem 1 If x_1 , x_2 are two eigenvectors of A corresponding to the eigenvalue λ_0 , then $k_1x_1 + k_2x_2$ is also an eigenvector for λ_0 , where k_1 , k_2 are any numbers that make $k_1x_1 + k_2x_2 \neq 0$.

定理1 若 x_1 , x_2 是 A 属于 λ_0 的两个特征向量,则 $k_1x_1 + k_2x_2$ 也是 A 属于 λ_0 的特征向量 (其中 k_1 , k_2 是任意常数,但 $k_1x_1 + k_2x_2 \neq \mathbf{0}$).

Proof x_1, x_2 are solutions to the following homogeneous system of linear equations:

$$(A - \lambda_0 I) x = 0,$$

So $k_1 x_1 + k_2 x_2$ is also a solution.

Therefore, nonzero vector $k_1 \mathbf{x}_1 + k_2 \mathbf{x}_2$ is also an eigenvector of \mathbf{A} corresponding to the eigenvalue λ_0 .

Theorem 2 Let $A = [a_{ij}]_{n \times n}$, and $\lambda_1, \lambda_2, \dots, \lambda_n$ be the n eigenvalues of A.

Then

(1)
$$\prod_{i=1}^{n} \lambda_i = |A| \text{ (i.e., det } A),$$

(2)
$$\sum_{i=1}^{n} \lambda_i = \sum_{i=1}^{n} a_{ii} = \text{trace}(A)(\text{i.e., tr}(A)).$$

The sum of all diagonal entries of A is called the trace of A (A 的迹).

*Some properties about the traces of matrices:

$$\operatorname{tr}(\boldsymbol{A}_{n\times n}+\boldsymbol{B}_{n\times n})=\operatorname{tr}(\boldsymbol{A}_{n\times n})+\operatorname{tr}(\boldsymbol{B}_{n\times n}), \quad \operatorname{tr}(\boldsymbol{A}_{m\times n}\boldsymbol{B}_{n\times m})=\operatorname{tr}(\boldsymbol{B}_{n\times m}\boldsymbol{A}_{m\times n}).$$

Proof

(1) Let $\lambda_1, \lambda_2, \dots, \lambda_n$ be the eigenvalues of the square matrix A.

Then the characteristic polynomial of A can be expressed as

$$|A - \lambda I| = (\lambda_1 - \lambda)(\lambda_2 - \lambda) \cdots (\lambda_n - \lambda)$$

Let $\lambda = 0$ on both sides, we can immediately get

$$|A| = \prod_{i=1}^{n} \lambda_{i}.$$

(make a clever choice of λ)

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

Among the expansion of the determinant, there is a term $(a_{11} - \lambda)(a_{22} - \lambda)\cdots(a_{nn} - \lambda)$, and the terms λ^n, λ^{n-1} in the characteristic polynomial *only* come from this term.

Therefore, $|A - \lambda I|$

$$= (-\lambda)^n + (a_{11} + a_{22} + \dots + a_{nn})(-\lambda)^{n-1} + \dots + |A|.$$

On the other hand

$$|A - \lambda I| = (\lambda_1 - \lambda)(\lambda_2 - \lambda) \cdots (\lambda_n - \lambda)$$

$$= (-\lambda)^n + (\lambda_1 + \lambda_2 + \cdots + \lambda_n)(-\lambda)^{n-1} \cdots + \prod_{i=1}^n \lambda_i,$$

$$\sum_{i=1}^n \lambda_i - \sum_{i=1}^n \alpha_i \quad \text{(find the coefficient of } \alpha_i)$$

Finally,
$$\sum_{i=1}^{n} \lambda_i = \sum_{i=1}^{n} a_{ii}.$$
 (find the coefficient of $(-\lambda)^{n-1}$ and compare)

Remark: Zero is an eigenvalue of A if and only if A is not invertible. 矩阵A可逆的充要条件是A的任意一个特征值不等于零.

A为奇异(singular)矩阵的充要条件是A至少有一个特征值等于零.

Note: A certain eigenvector of *A* cannot be corresponding to different eigenvalues. (*A*的一个特征向量不能属于不同的特征值.)

If x were an eigenvector of A corresponding to different eigenvalues λ_1 , λ_2 ($\lambda_1 \neq \lambda_2$),

i.e.,
$$A x = \lambda_1 x$$
 and $A x = \lambda_2 x$,

then
$$\lambda_1 x = \lambda_2 x$$
, that is, $(\lambda_1 - \lambda_2) x = 0$.

Since $\lambda_1 - \lambda_2 \neq 0$, then x = 0, which is impossible for an eigenvector.

Property 1 If $\lambda_1, \lambda_2, ..., \lambda_n$ are the eigenvalues of A, then $\lambda_1^k, \lambda_2^k, ..., \lambda_n^k$ are the eigenvalues of A^k , where k is a positive integer, and k may equal -1 if A is invertible.

Moreover, if x is an eigenvector of A corresponding to λ , then x is also an eigenvector of A^k corresponding to λ^k .

Proof. Notice that, if x is an eigenvector of A corresponding to λ_i , then

$$A^2x = A(Ax) = A(\lambda_i x) = \lambda_i Ax = \lambda_i (\lambda_i x) = \lambda_i^2 x$$

The proposition then follows.

性质1 若 λ 是A的特征值, x 是A的属于 λ 的特征向量. 则

- (1) $k\lambda$ 是kA的特征值(k为任意常数);
- (2) $\lambda^m \mathbf{A}^m$ 的特征值;
- (3) 若A可逆,则 λ^{-1} 为 A^{-1} 的一个特征值;

且 x 仍然是矩阵kA, A^m 和 A^{-1} 的分别对应于特征值 $k\lambda$, λ^m 和 λ^{-1} 的特征向量.

Example 5 Suppose a 3×3 matrix A has eigenvalues 1, -1, 2. And

$$B = A^3 - 5A^2$$
. Find $|B|$.

Solution The eigenvalues of **B** are: $1^3 - 5 \cdot 1^2 = -4$,

$$(-1)^3 - 5 \cdot (-1)^2 = -6,$$
 $2^3 - 5 \cdot 2^2 = -12.$

So
$$|\mathbf{B}| = (-4)(-6)(-12) = -288$$
.

Property 2 The matrices A and A^{T} have same eigenvalues.

(性质2 矩阵A和 A^{T} 的特征值相同.)

Proof. det
$$(A - \lambda I) = \det (A - \lambda I)^T$$

= det $(A^T - (\lambda I)^T) = \det (A^T - \lambda I)$.

Thus A and A^{T} have the same characteristic polynomials.

The proposition then follows.

回顾:人口流动问题

Possible? Yes!

$$A = PQP^{-1}$$

$$A^k = PQ^kP^{-1}$$
 (Q: diagonal)

$$AP = PQ$$

$$\begin{bmatrix} \mathbf{A} \\ 3 \end{bmatrix} = 1 \times \begin{vmatrix} 2 \\ 3 \end{vmatrix}, \quad \begin{bmatrix} \mathbf{A} \\ -1 \end{vmatrix} = \begin{bmatrix} 7 \\ 12 \end{vmatrix} = 1 \end{vmatrix}$$

$$\mathbf{A} = \begin{bmatrix} \frac{3}{4} & \frac{1}{6} \\ \frac{1}{4} & \frac{5}{6} \end{bmatrix}$$

$$Q = \begin{bmatrix} 1 & & \\ & \frac{7}{12} \end{bmatrix} \qquad P = \begin{bmatrix} 2 & 1 \\ 3 & -1 \end{bmatrix}$$

$$\mathbf{P} = \begin{vmatrix} 2 & 1 \\ 3 & -1 \end{vmatrix}$$

$$\begin{bmatrix} y_k \\ z_k \end{bmatrix} = \mathbf{A}^k \begin{bmatrix} y_0 \\ z_0 \end{bmatrix} = \mathbf{P} \mathbf{Q}^k \mathbf{P}^{-1} \begin{bmatrix} y_0 \\ z_0 \end{bmatrix}$$

$$= \begin{bmatrix} \frac{2}{5} (y_0 + z_0) + \frac{1}{5} \times \left(\frac{7}{12}\right)^k (3y_0 - 2z_0) \\ \frac{3}{5} (y_0 + z_0) + \frac{1}{5} \times \left(\frac{7}{12}\right)^k (2z_0 - 3y_0) \end{bmatrix} \rightarrow \begin{bmatrix} \frac{2}{5} (y_0 + z_0) \\ \frac{3}{5} (y_0 + z_0) \end{bmatrix}, \quad \stackrel{\text{def}}{=} k \rightarrow \infty \text{ Bf}.$$

农业与非农业人口比例的趋势为 2:3.



- ⋄ y₀与z₀的比例对稳定的趋势有多大影响?
- ♠ 稳定的趋势与A的什么特征有联系?
- 如果人口总数是变化的, 如何建模?



$$A = \begin{bmatrix} \frac{3}{4} & \frac{1}{6} \\ \frac{1}{4} & \frac{5}{6} \end{bmatrix}.$$

$$= PQP^{-1}.$$

$$\begin{bmatrix} 2 & 1 \\ 3 & -1 \end{bmatrix} \begin{bmatrix} 1 \\ 7/12 \end{bmatrix}$$

对 $A_{n\times n}$,

- 不同的特征值对应的 特征向量线性无关吗?
- 是否存在可逆矩阵 P
 和对角矩阵 Q, 使得
 A=PQP-1 ?

Come back later in 5.2:

Diagonalization (对角化)

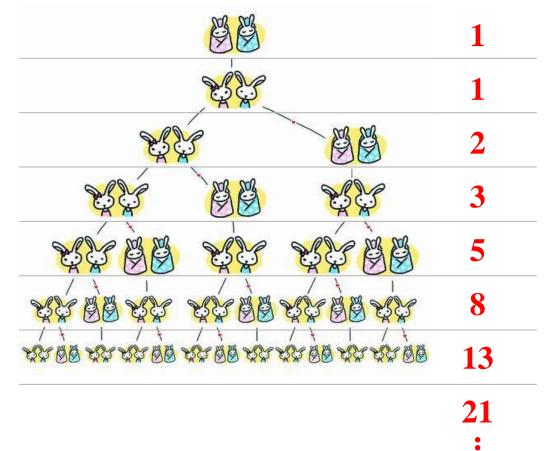
of a Matrix.

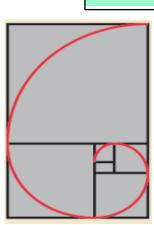
Another interesting application...

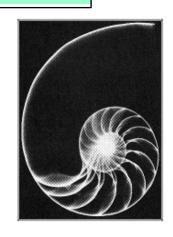
用矩阵方法求Fibonacci数列的通项公式



Fibonacci 1175-1250







1

Fibonacci数列的递推关系为

$$F_0=0, F_1=1, F_{n+2}=F_{n+1}+F_n, n=0,1,2,...$$

首先构造一组恒等式

$$\begin{cases}
F_{n+2} = F_{n+1} + F_n, \\
F_{n+1} = F_{n+1}.
\end{cases} \qquad \qquad \qquad \begin{bmatrix}
F_{n+2} \\
F_{n+1}
\end{bmatrix} = \begin{bmatrix}
1 & 1 \\
1 & 0
\end{bmatrix} \begin{bmatrix}
F_{n+1} \\
F_n
\end{bmatrix}.$$

$$A \qquad u_n$$

A的特征值为

$$\lambda_1 = \frac{1 - \sqrt{5}}{2}, \lambda_2 = \frac{1 + \sqrt{5}}{2}.$$

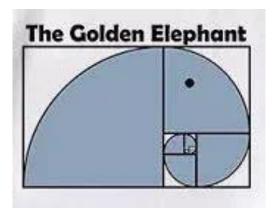
进而求出数列通项

$$F_n = \frac{\sqrt{5}}{5} \left[\left(\frac{1 + \sqrt{5}}{2} \right)^n - \left(\frac{1 - \sqrt{5}}{2} \right)^n \right].$$

 $|\boldsymbol{u}_n = \boldsymbol{A}\boldsymbol{u}_{n-1} = \boldsymbol{A}^n\boldsymbol{u}_0.$

Key words:

Definition
Calculation
Properties
Examples



Homework

See Blackboard

