

Linear Algebra



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5

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

5.1

EIGENVALUES AND EIGENVECTORS

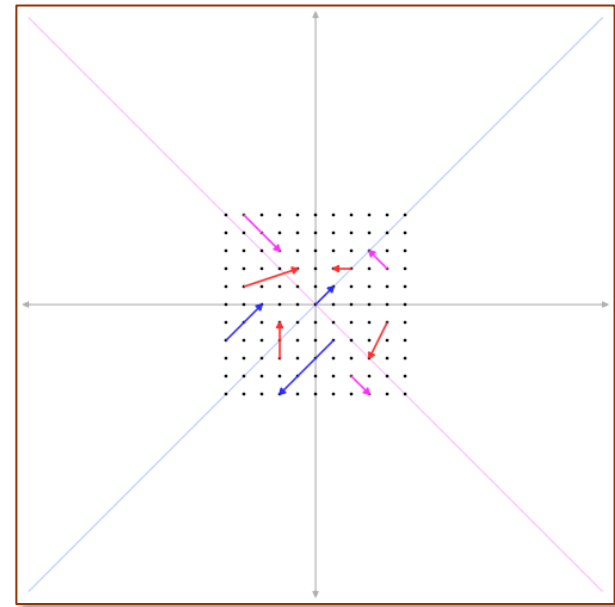
Introduction

Definition

Calculations

Properties

Application



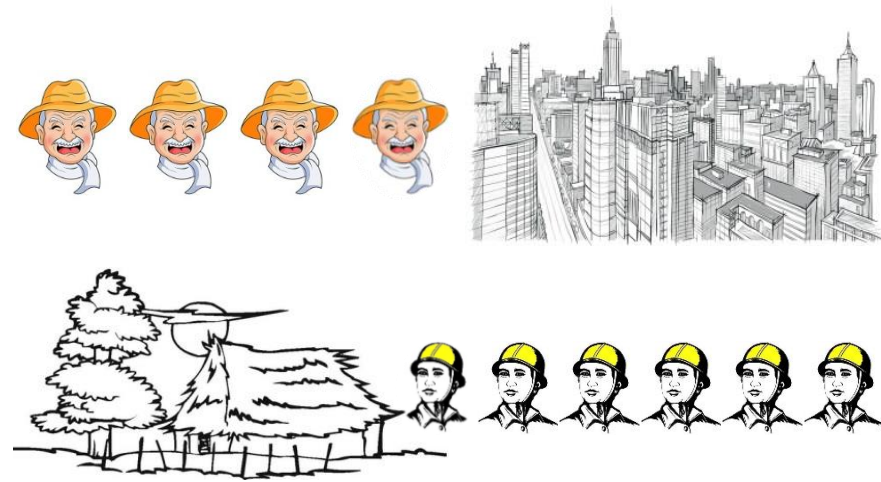
I. Introductory example - 人口流动问题



简化模型

❶ 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 ,

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

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 ,

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

矩阵 A

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 ,

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

矩阵 A

如何计算
方阵 A 的幂
 A^k ?

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

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

有无可能?

设想

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



$$A^k = PQ^k P^{-1}$$

(其中 Q 为对角阵)

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

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

$$AP = PQ$$



$$A[P_1 \mid P_2] = [P_1 \mid P_2] \begin{bmatrix} \lambda_1 \\ \lambda_2 \end{bmatrix}$$

(其中 P_1, P_2 为2维列向量)

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



$$[AP_1 \mid AP_2] = [\lambda_1 P_1 \mid \lambda_2 P_2]$$

$$A_{2 \times 2} \mathbf{x}_{2 \times 1} = \lambda \mathbf{x}_{2 \times 1}$$

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

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

then λ is called an **eigenvalue (特征值)** of A , and \mathbf{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**?

$$R^n \rightarrow R^n$$

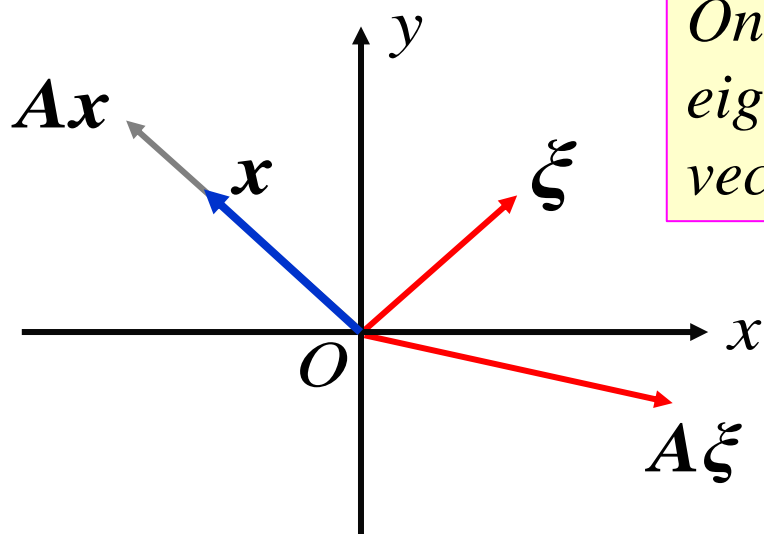
$$A : \mathbf{x} \rightarrow A\mathbf{x}$$

$$R^n \rightarrow R^n$$

$$\lambda : \mathbf{x} \rightarrow \lambda \mathbf{x}$$

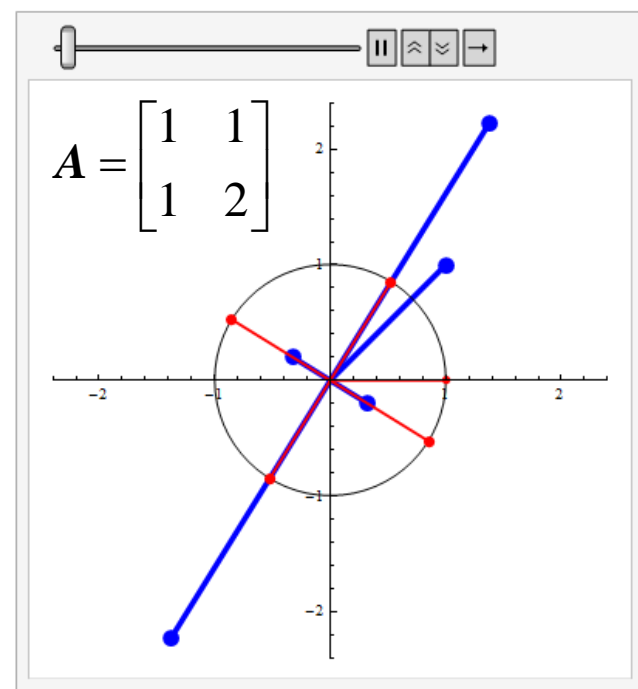
给定方阵 $A \in \mathbf{R}^{n \times n}$, 一般说来, 对于 $\xi \in \mathbf{R}^n$, 向量 $A\xi$ 与 ξ 不在 同一方向上.

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

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

which shows that 4 is an eigenvalue of \mathbf{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 \mathbf{A} , and $(1, -2/3)^T$ is an eigenvector corresponding to -1 .

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

How to get?

!dea

Equivalently, an eigenvalue λ and a corresponding eigenvector \mathbf{x} satisfy:

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

Since \mathbf{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 \mathbf{x} 是非零向量, 是齐次线性方程组

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

的非零解. λ 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

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

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.

(λ 是 A 的特征值, 当且仅当 λ 是特征多项式的根.)

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 $n \geq 5$ 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 \mathbf{A} can be obtained as follows.

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

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

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

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

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

It follows that $\mathbf{x}_2 = k_2(5,2)^T$ ($k_2 \in \mathbf{R}$, $k_2 \neq 0$) 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 $(A - \lambda I) \mathbf{x} = \mathbf{0}$.*

Since the determinant is zero, there are solutions other than $\mathbf{x} = \mathbf{0}$. Those are the eigenvectors.

特征值与特征向量的求解 步骤

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.

$$A \xrightarrow{\text{求特征值}} |A - \lambda I| = 0 \xrightarrow{\text{求特征向量}} (A - \lambda_i I)x = 0$$

求特征值 λ_i

求特征向量

Remark: 另一种等价求法

$$A \xrightarrow{\text{求特征值}} |\lambda I - A| = 0 \xrightarrow{\text{求特征向量}} (\lambda_i I - A)x = 0$$

求特征值 λ_i

求特征向量

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

$$A = \begin{bmatrix} 3 & 0 \\ 0 & 2 \end{bmatrix} \text{ has } \lambda_1 = 3 \text{ with } \mathbf{x}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix},$$

$$\lambda_2 = 2 \text{ with } \mathbf{x}_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}.$$

On each eigenvector A acts like: $A\mathbf{x}_1 = 3\mathbf{x}_1$ and $A\mathbf{x}_2 = 2\mathbf{x}_2$.

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:
$$A(\mathbf{x}_1 + 5\mathbf{x}_2) = 3\mathbf{x}_1 + 10\mathbf{x}_2 = \begin{bmatrix} 3 \\ 10 \end{bmatrix}.$$

This is $A\mathbf{x}$ for a typical vector \mathbf{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*.

(n 阶对角矩阵 A 的特征值是它的 n 个主对角元 $a_{11}, a_{22}, \dots, a_{nn}$.)

This is true since the characteristic polynomial of A is

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

2. The eigenvalues are on the main diagonal when B is *triangular*.

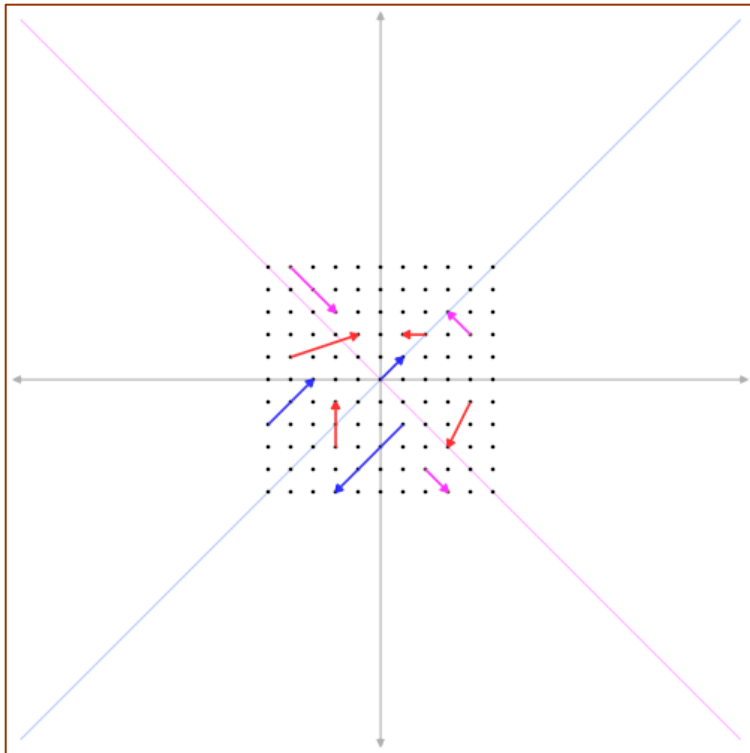
(n 阶上(下)三角形矩阵 B 的特征值也是它的 n 个主对角元 $b_{11}, b_{22}, \dots, b_{nn}$.)

Exercise

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

Example 3 Find the eigenvalues and eigenvectors of the matrix

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



The Eigenvectors

$$k_1 \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

$$(k_1 \neq 0)$$

$$k_2 \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$$(k_2 \neq 0)$$

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

$$A^2\mathbf{x} = A(A\mathbf{x}) = A(\lambda\mathbf{x}) = \lambda A\mathbf{x} = \lambda\lambda\mathbf{x} = \lambda^2\mathbf{x}$$

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

Since $\mathbf{x} \neq \mathbf{0}$, 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.

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

$$|A_1 - \lambda I| = -\lambda(\lambda + 2)^2$$

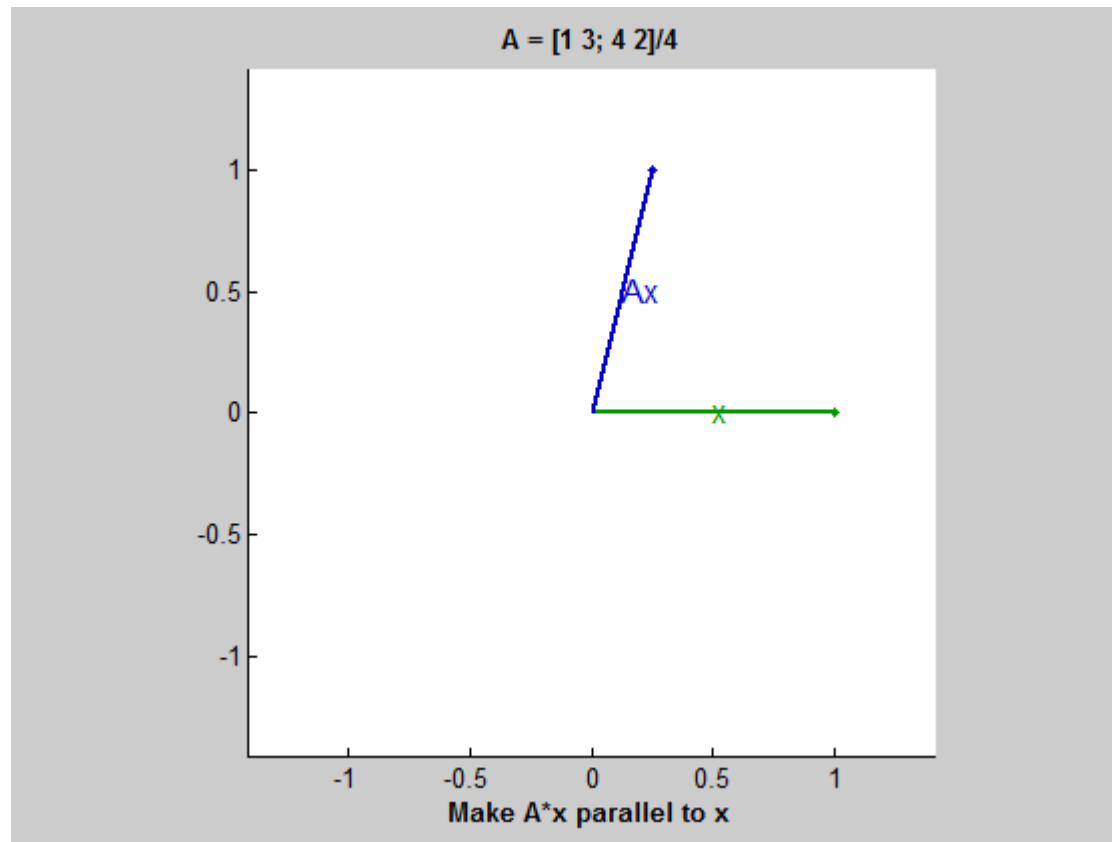
$$A_2 = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} \text{ has eigenvalues: } \lambda = \cos \theta \pm i \sin \theta.$$

$$|A_2 - \lambda I| = \lambda^2 - (2 \cos \theta)\lambda + 1$$

Matlab demo: eigshow

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

$$\lambda_1 = -0.5, \\ \lambda_2 = 1.25.$$

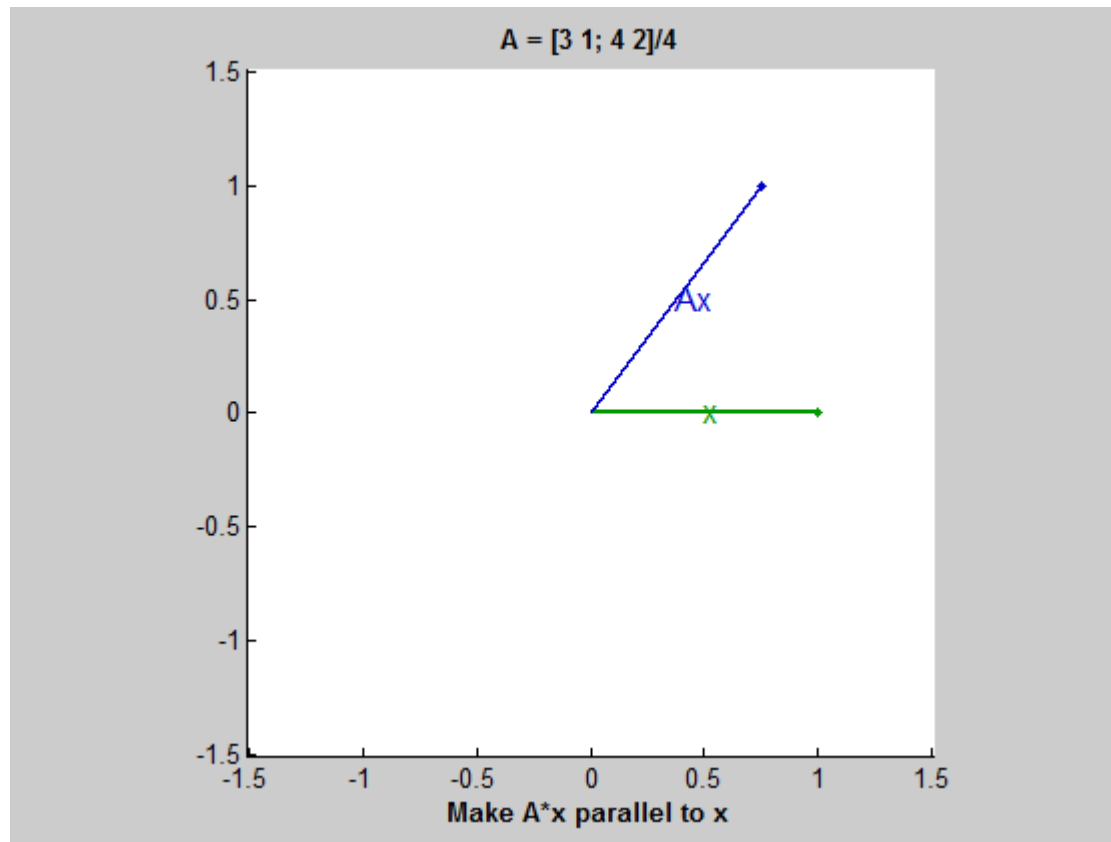


<https://blogs.mathworks.com/cleve/2013/07/08/eigshow-week-1/#f96996aa-ef86-4137-b343-07584baff36c>

Matlab demo: eigshow

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

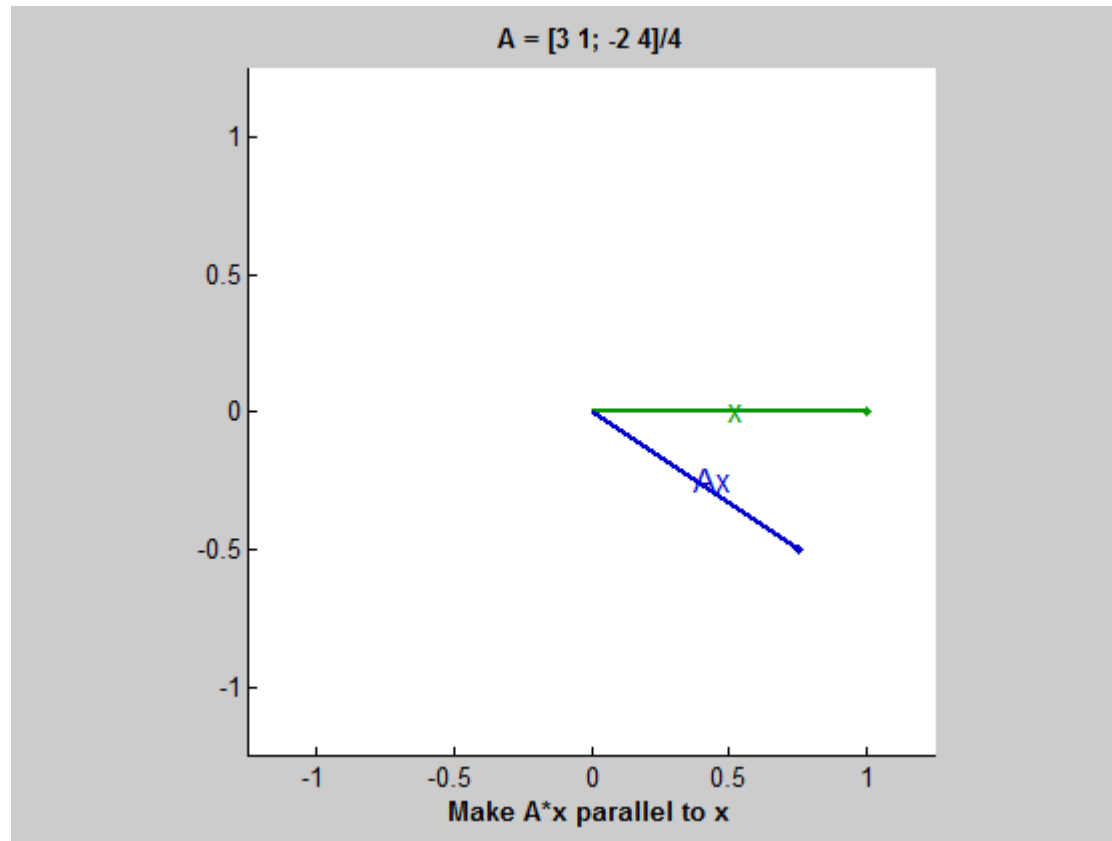
$$\lambda_1 = 1.14,$$
$$\lambda_2 = 0.11.$$



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

Matlab demo: eigshow

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



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

III. Eigenvalues and Eigenvectors – Properties

Theorem 1 If $\mathbf{x}_1, \mathbf{x}_2$ are two eigenvectors of A corresponding to the eigenvalue λ_0 , then $k_1\mathbf{x}_1 + k_2\mathbf{x}_2$ is also an eigenvector for λ_0 , where k_1, k_2 are any numbers that make $k_1\mathbf{x}_1 + k_2\mathbf{x}_2 \neq \mathbf{0}$.

定理1 若 $\mathbf{x}_1, \mathbf{x}_2$ 是 A 属于 λ_0 的两个的特征向量, 则 $k_1\mathbf{x}_1 + k_2\mathbf{x}_2$ 也是 A 属于 λ_0 的特征向量 (其中 k_1, k_2 是任意常数, 但 $k_1\mathbf{x}_1 + k_2\mathbf{x}_2 \neq \mathbf{0}$).

Proof $\mathbf{x}_1, \mathbf{x}_2$ are solutions to the following homogeneous system of linear equations:

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

So $k_1\mathbf{x}_1 + k_2\mathbf{x}_2$ is also a solution.

Therefore, nonzero vector $k_1\mathbf{x}_1 + k_2\mathbf{x}_2$ is also an eigenvector of 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

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

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

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

*Some properties about the traces of matrices:

$$\text{tr}(A_{n \times n} + B_{n \times n}) = \text{tr}(A_{n \times n}) + \text{tr}(B_{n \times n}), \quad \text{tr}(A_{m \times n} B_{n \times m}) = \text{tr}(B_{n \times m} A_{m \times n}).$$

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)

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

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.

$$\begin{aligned} \text{Therefore, } |\mathbf{A} - \lambda \mathbf{I}| &= (-\lambda)^n + (a_{11} + a_{22} + \cdots + a_{nn})(-\lambda)^{n-1} + \cdots + |\mathbf{A}|. \end{aligned}$$

On the other hand

$$\begin{aligned} |\mathbf{A} - \lambda \mathbf{I}| &= (\lambda_1 - \lambda)(\lambda_2 - \lambda) \cdots (\lambda_n - \lambda) \\ &= (-\lambda)^n + (\lambda_1 + \lambda_2 + \cdots + \lambda_n)(-\lambda)^{n-1} + \cdots + \prod_{i=1}^n \lambda_i, \end{aligned}$$

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

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 \mathbf{x} were an eigenvector of A corresponding to different eigenvalues $\lambda_1, \lambda_2 (\lambda_1 \neq \lambda_2)$,

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

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

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

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

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

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

$$\mathbf{A}^2 \mathbf{x} = \mathbf{A}(\mathbf{A} \mathbf{x}) = \mathbf{A}(\lambda_i \mathbf{x}) = \lambda_i \mathbf{A} \mathbf{x} = \lambda_i (\lambda_i \mathbf{x}) = \lambda_i^2 \mathbf{x},$$

The proposition then follows.

性质1 若 λ 是 \mathbf{A} 的特征值, \mathbf{x} 是 \mathbf{A} 的属于 λ 的特征向量. 则

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

且 \mathbf{x} 仍然是矩阵 $k\mathbf{A}$, \mathbf{A}^m 和 \mathbf{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 $|B| = (-4)(-6)(-12) = -288$.

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!

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

$$A^k = P Q^k P^{-1} \quad (Q: \text{diagonal})$$

$$AP = PQ$$

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

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

$$Q = \begin{bmatrix} 1 & \\ & \frac{7}{12} \end{bmatrix}$$

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

$$\begin{aligned}
\begin{bmatrix} y_k \\ z_k \end{bmatrix} &= \mathbf{A}^k \begin{bmatrix} y_0 \\ z_0 \end{bmatrix} = \mathbf{P} \mathbf{Q}^k \mathbf{P}^{-1} \begin{bmatrix} y_0 \\ z_0 \end{bmatrix} \\
&= \begin{bmatrix} \frac{2}{5}(y_0 + z_0) + \frac{1}{5} \times \left(\frac{7}{12}\right)^k (3y_0 - 2z_0) \\ \frac{3}{5}(y_0 + z_0) + \frac{1}{5} \times \left(\frac{7}{12}\right)^k (2z_0 - 3y_0) \end{bmatrix} \rightarrow \begin{bmatrix} \frac{2}{5}(y_0 + z_0) \\ \frac{3}{5}(y_0 + z_0) \end{bmatrix}, \quad \text{当 } k \rightarrow \infty \text{ 时.}
\end{aligned}$$

农业与非农业人口比例的趋势为 2 : 3.



- ❶ y_0 与 z_0 的比例对稳定的趋势有多大影响?
- ❷ 稳定的趋势与 \mathbf{A} 的什么特征有联系?
- ❸ 如果人口总数是变化的, 如何建模?



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

$$= \mathbf{P} \mathbf{Q} \mathbf{P}^{-1}.$$

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

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

$$A = P Q P^{-1} ?$$

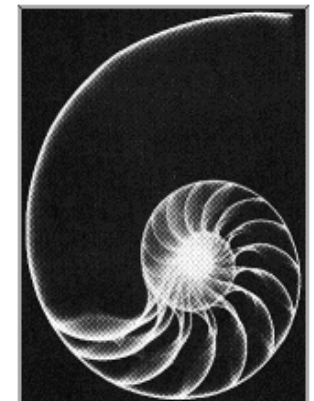
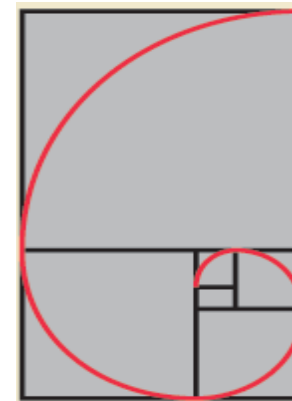
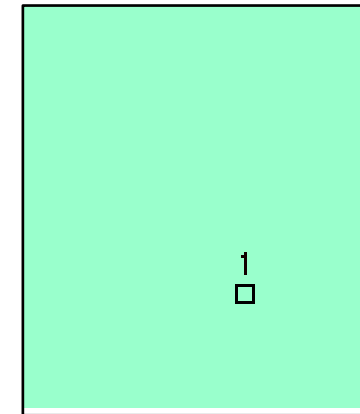
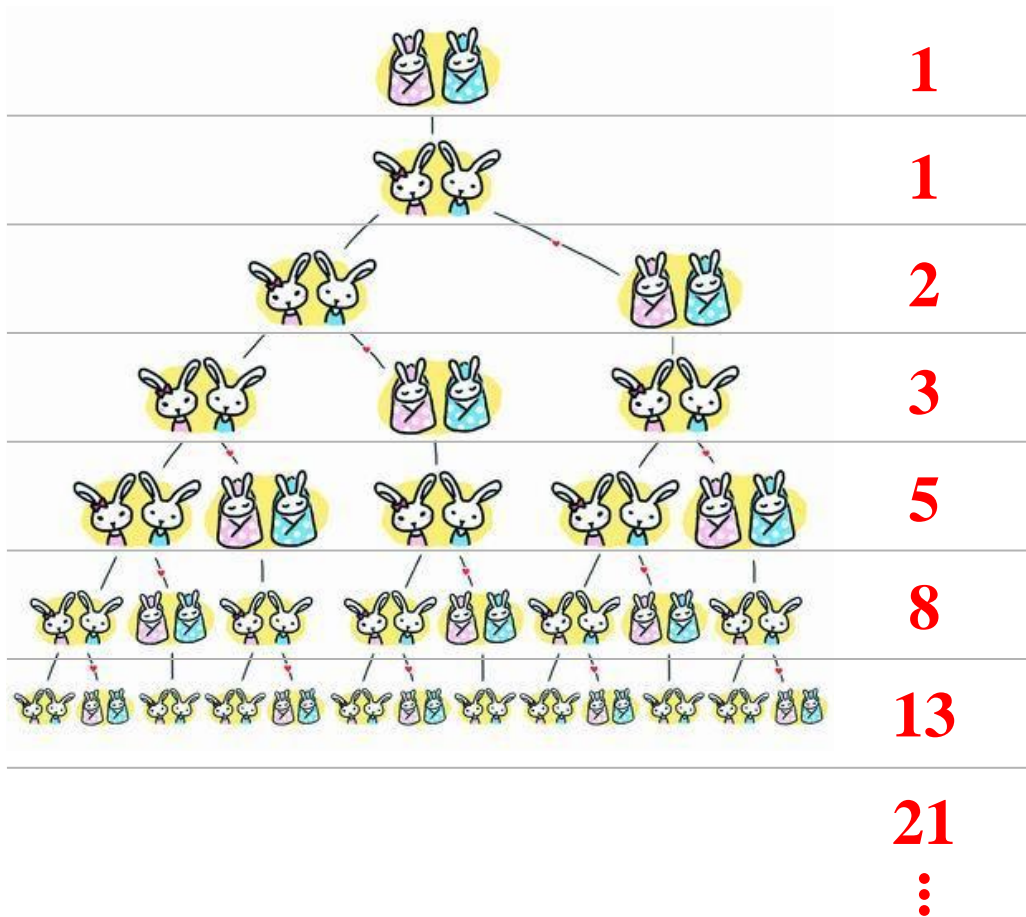
Come back later in 5.2:
Diagonalization (对角化)
of a Matrix.

Another interesting application...

用矩阵方法求 *Fibonacci* 数列的通项公式



Fibonacci
1175-1250



*Fibonacci*数列的递推关系为

$$F_0=0, F_1=1, F_{n+2}=F_{n+1}+F_n, \quad n=0,1,2,\dots$$

首先构造一组恒等式

$$\begin{cases} F_{n+2} = F_{n+1} + F_n, \\ F_{n+1} = F_{n+1}. \end{cases} \quad \Rightarrow \quad \begin{bmatrix} F_{n+2} \\ F_{n+1} \end{bmatrix} = \underbrace{\begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}}_{\mathbf{A}} \underbrace{\begin{bmatrix} F_{n+1} \\ F_n \end{bmatrix}}_{\mathbf{u}_n}.$$

$$\Rightarrow \quad \mathbf{u}_n = \mathbf{A}\mathbf{u}_{n-1} = \mathbf{A}^n \mathbf{u}_0.$$

\mathbf{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

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

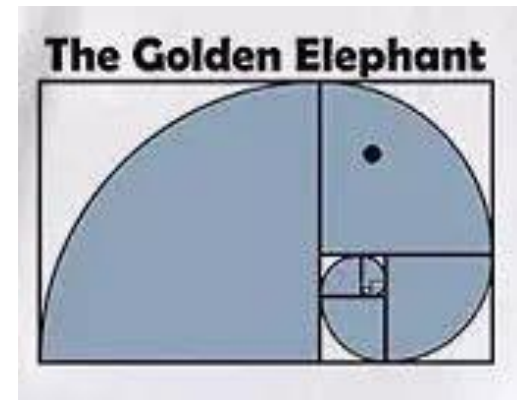
Key words:

Definition

Calculation

Properties

Examples



Homework

See Blackboard

