

LECTURE 4

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SEP 18TH 2012

GRA 6035

MATHEMATICS

REVIEW OF LECTURE 3:

The vectors $\{\underline{v}_1, \underline{v}_2, \dots, \underline{v}_n\} \in \mathbb{R}^m$ are linearly independent if

$$c_1 \underline{v}_1 + c_2 \underline{v}_2 + \dots + c_n \underline{v}_n = \underline{0}$$

has only the trivial solution $c_1 = c_2 = \dots = c_n = 0$. This will be the case if and only if

$$\text{rk} \left(\begin{array}{c|c|c} \underline{v}_1 & \underline{v}_2 & \dots & \underline{v}_n \end{array} \right) = n$$

Ex1

$$A = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{pmatrix} \quad \underline{v}_1 = \begin{pmatrix} 1 \\ 4 \end{pmatrix} \quad \underline{v}_2 = \begin{pmatrix} 2 \\ 5 \end{pmatrix} \quad \underline{v}_3 = \begin{pmatrix} 3 \\ 6 \end{pmatrix}$$

$\text{rk } A = 2$ since $\begin{vmatrix} 1 & 2 \\ 4 & 5 \end{vmatrix} = 5 - 8 = -3 \neq 0$, or alternatively

$$\rightarrow \begin{array}{c} \text{R}_2 - 4\text{R}_1 \\ \text{R}_3 - 3\text{R}_1 \end{array} \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 0 & -3 & -6 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 2 & 3 \\ 0 & -3 & -6 \\ 0 & -3 & -6 \end{pmatrix} \begin{array}{l} c_1 + 2c_2 + 3c_3 = 0 \\ -3c_2 - 6c_3 = 0 \end{array}$$

Conclusion: $\{\underline{v}_1, \underline{v}_2, \underline{v}_3\}$ not linearly independent

$\{\underline{v}_1, \underline{v}_2\}$

are

and

$$1 \cdot \underline{v}_1 + 2 \underline{v}_2 + 1 \cdot \underline{v}_3 = \underline{0} \Rightarrow \underline{v}_3 = -\underline{v}_1 - 2\underline{v}_2$$

$$\begin{array}{l} c_1 = c_3 \\ c_2 = -2c_3 \\ c_3 \text{ free} \end{array}$$

$$c_1 = 1, c_2 = -2, c_3 = 1$$

OFFICE HOURS:

New office hours:

Mon 12-14

at B4-032 (my office).

Plan Lecture 4

- ① Eigenvalues, eigenvectors
- ② Diagonalization

Reading:

[FMEAT] 1.5-1.6

① Eigenvalues and eigenvectors

A : $n \times n$ -matrix

Defn: An eigenvalue of A is a number λ such that

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

has a non-zero solution $\underline{x} \neq \underline{0}$, (a vector)

If λ is an eigenvalue for A , then all the vectors $\underline{x} \in \mathbb{R}^n$ (\underline{x} is an n -vector) such that

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

Ex: $A = \begin{pmatrix} 2 & 1 \\ 1 & 2 \end{pmatrix}$

$\underline{x} = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$: $A\underline{x} = \begin{pmatrix} 2 & 1 \\ 1 & 2 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 1 \end{pmatrix} = \begin{pmatrix} 3 \\ 3 \end{pmatrix}$

$\lambda \cdot \underline{x} = \lambda \cdot \begin{pmatrix} 1 \\ 1 \end{pmatrix} = \begin{pmatrix} \lambda \\ \lambda \end{pmatrix}$

Conclusion: $\lambda = 3$, $\underline{x} = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$

is a solution

$\lambda = 3$ eigenvalue

$\underline{x} = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$ eigenvector

$$A = \begin{pmatrix} 2 & 1 \\ 1 & 2 \end{pmatrix}; \quad \underline{x} = \begin{pmatrix} 1 \\ 2 \end{pmatrix}$$

$$A\underline{x} = \begin{pmatrix} 2 & 1 \\ 1 & 2 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 2 \end{pmatrix} = \begin{pmatrix} 4 \\ 5 \end{pmatrix}$$

$$\lambda \underline{x} = \lambda \cdot \begin{pmatrix} 1 \\ 2 \end{pmatrix} = \begin{pmatrix} \lambda \\ 2\lambda \end{pmatrix}$$

no solutions of $A\underline{x} = \lambda \underline{x}$

$$\begin{pmatrix} 4 \\ 5 \end{pmatrix} = \begin{pmatrix} \lambda \\ 2\lambda \end{pmatrix} \quad \begin{array}{l} \lambda = 4 \\ 2\lambda = 5 \end{array}$$

$8 = 5$ impossible.

$\underline{x} = \begin{pmatrix} 1 \\ 2 \end{pmatrix}$ is not eigenvector

Standard method: Finding eigenvalues and eigenvectors for A.

① Find eigenvalues

$A\underline{x} = \lambda \underline{x}$ has non-trivial solutions

$$A\underline{x} - \lambda \underline{x} = \underline{0} \quad \text{---} \quad \text{||} \quad \text{---}$$

$n \times n$ matrix \rightarrow $(A - \lambda I)\underline{x} = \underline{0} \quad \text{---} \quad \text{||} \quad \text{---}$

$$\boxed{\det(A - \lambda I) = 0}$$

Characteristic equation

Ex: $A = \begin{pmatrix} 2 & 1 \\ 1 & 2 \end{pmatrix}$

Char. eqn: $\det(A - \lambda I) = 0$

$$\det\left(\begin{pmatrix} 2 & 1 \\ 1 & 2 \end{pmatrix} - \lambda \cdot \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}\right) = 0$$

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

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

$$4 - 4\lambda + \lambda^2 - 1 = 0$$

$$\lambda^2 - 4\lambda + 3 = 0$$

$$\underline{\lambda = 3, \lambda = 1}$$

If we look at the linear system

$$C \cdot \underline{x} = \underline{0}$$

$\det C = 0$: free variables non-zero sol's.

$\det C \neq 0$: $C^{-1} \cdot C\underline{x} = C^{-1} \cdot \underline{0}$

$$\underline{x} = \underline{0}$$

Conclusions:

Two eigenvalues for A:

$$\underline{\lambda_1 = 3, \lambda_2 = 1}$$

Facts:

- ① A $n \times n$ -matrix $\Rightarrow |A - \lambda I| = 0$ is an n 'th order polynomial eqn. in λ .
- ② $A = \begin{pmatrix} a & b \\ c & d \end{pmatrix} \Rightarrow |A - \lambda I| = 0$ can be written as
 $\lambda^2 - \text{tr} A \cdot \lambda + \det(A) = 0$
($\text{tr} A = a + d$, $\det A = ad - bc$)
- ③ If A is a symmetric matrix, then:
 $|A - \lambda I| = (-1)^n \cdot (\lambda - \lambda_1)(\lambda - \lambda_2) \dots (\lambda - \lambda_n)$
where $\lambda_1, \lambda_2, \dots, \lambda_n$ are the eigenvalues.
- ④ If A has n eigenvalues $\lambda_1, \lambda_2, \dots, \lambda_n$, then:
 $\det(A) = \lambda_1 \cdot \lambda_2 \cdot \lambda_3 \dots \lambda_n$
 $\text{tr}(A) = \lambda_1 + \lambda_2 + \lambda_3 + \dots + \lambda_n$
-

Ex: $A = \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix}$
non-symmetric

$$|A - \lambda I| = \begin{vmatrix} -\lambda & 1 \\ -1 & -\lambda \end{vmatrix} = \lambda^2 + 1 = 0$$
$$\lambda^2 + 1 = 0 \quad \lambda^2 = -1$$

No eigenvalues

Ex: $A = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 1 \end{pmatrix}$
diagonal

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

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

$$\lambda_1 = 1, \lambda_2 = 2, \lambda_3 = 1$$

$$\lambda = 1 \quad \text{mult. } 2$$
$$\lambda = 2 \quad \text{mult. } 1$$

$$\det(A) = 1 \cdot 2 \cdot 1 = 2$$

$$\text{tr}(A) = 1 + 2 + 1 = 4$$

Ex: $A = \begin{pmatrix} 2 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 2 \end{pmatrix}$

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

$$(1-\lambda) \cdot \begin{vmatrix} 2-\lambda & 1 \\ 1 & 2-\lambda \end{vmatrix} = 0$$

$$(1-\lambda) \cdot (\lambda^2 - 4\lambda + 3) = 0$$

$$\underline{\lambda_1 = 1} \quad \underline{\lambda_2 = 3}, \underline{\lambda_3 = 1}$$

Standard method: Eigenvectors; $A\underline{x} = \lambda\underline{x}$

② If λ is an eigenvalue: (λ found in ①)

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

$$\underline{(A - \lambda I)\underline{x} = \underline{0}} \leftarrow \text{linear system; we solve it for } \underline{x}.$$

Ex: $A = \begin{pmatrix} 2 & 1 \\ 1 & 2 \end{pmatrix}$ In step ① we found that $\underline{\lambda_1 = 3}$ and $\underline{\lambda_2 = 1}$ are the eigenvalues.

$$\underline{\lambda = 3}: \begin{pmatrix} 2-3 & 1 \\ 1 & 2-3 \end{pmatrix} \underline{x} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$\left[\begin{array}{cc|c} -1 & 1 & 0 \\ 1 & -1 & 0 \end{array} \right] \begin{matrix} \underline{x} \\ \underline{y} \end{matrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \quad \left[\begin{array}{cc|c} -1 & 1 & 0 \\ 1 & -1 & 0 \end{array} \right] \xrightarrow{+} \left[\begin{array}{cc|c} -1 & 1 & 0 \\ 0 & 0 & 0 \end{array} \right]$$

$$\underline{x} = \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} y \\ y \end{pmatrix} = \underline{y \cdot \begin{pmatrix} 1 \\ 1 \end{pmatrix}} \leftarrow \begin{matrix} -x+y=0 \Rightarrow x=y \\ y \text{ free} \end{matrix}$$

Eigenvectors for $\lambda=3$:

$$\underline{y \cdot \begin{pmatrix} 1 \\ 1 \end{pmatrix}}, \quad y \text{ free}$$

Eigenvectors for $\lambda=1$:

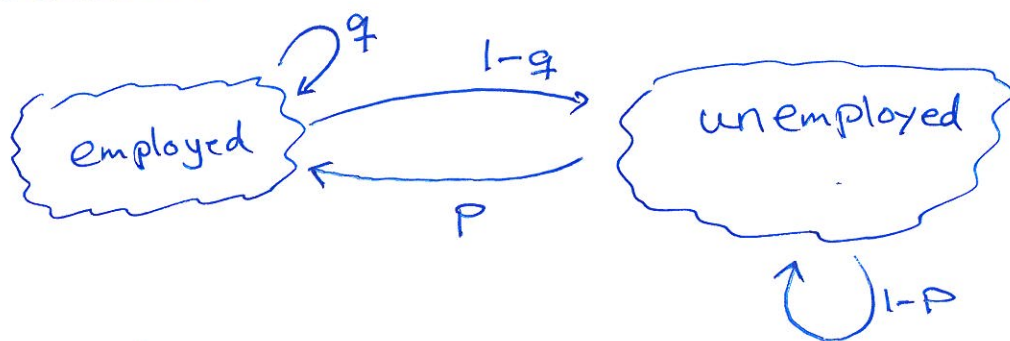
$$\begin{pmatrix} 2-1 & 1 \\ 1 & 2-1 \end{pmatrix} \cdot \underline{x} = \underline{0} \quad \leftarrow \quad (A - \lambda I) \underline{x} = \underline{0}$$

$$\begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \quad \begin{matrix} x+y=0 \\ y \text{ free} \end{matrix} \quad \underline{x} = \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} -y \\ y \end{pmatrix} = \underline{y} \cdot \underline{\underline{\begin{pmatrix} -1 \\ 1 \end{pmatrix}}}$$

Facts:

- ① For each eigenvalue λ for A , the equation $(A - \lambda I) \underline{x} = \underline{0}$ is a linear system with at least one degree of freedom.
- ② If λ has multiplicity m , then $(A - \lambda I) \underline{x} = \underline{0}$ has at most m degrees of freedom.

Motivation: Linear dynamical system



$$\underline{v}_0 = \begin{pmatrix} e_0 \\ u_0 \end{pmatrix} \quad \underline{v}_1 = \begin{pmatrix} e_1 \\ u_1 \end{pmatrix} = \begin{pmatrix} q & p \\ 1-q & 1-p \end{pmatrix} \cdot \begin{pmatrix} e_0 \\ u_0 \end{pmatrix} = A \cdot \underline{v}_0$$

↑
state
vector

$$\begin{aligned} e_1 &= q e_0 + p u_0 \\ u_1 &= (1-q) e_0 + (1-p) u_0 \end{aligned}$$

$$A = \begin{pmatrix} q & p \\ 1-q & 1-p \end{pmatrix}$$

transition matrix

$$\underline{v}_1 = A \underline{v}_0$$

$$\underline{v}_2 = A \cdot \underline{v}_1 = A \cdot A \underline{v}_0 = A^2 \underline{v}_0$$

⋮

$$\underline{v}_T = A^T \underline{v}_0$$

Ex: $\underline{v}_0 = \begin{pmatrix} 0.9 \\ 0.1 \end{pmatrix}$ $A = \begin{pmatrix} 0.998 & 0.136 \\ 0.002 & 0.864 \end{pmatrix}$

$$\underline{v}_1 = A \cdot \underline{v}_0 = \begin{pmatrix} 0.998 & 0.136 \\ 0.002 & 0.864 \end{pmatrix} \cdot \begin{pmatrix} 0.9 \\ 0.1 \end{pmatrix} = ?$$

$$\underline{v}_2 = A^2 \underline{v}_0 = \begin{pmatrix} 0.998 & \dots \\ \dots & \dots \end{pmatrix}^2 \cdot \begin{pmatrix} 0.9 \\ 0.1 \end{pmatrix} = ?$$

$$\vdots$$

$$\underline{v}_{30} = A^{30} \underline{v}_0 = \begin{pmatrix} 0.998 & \dots \\ \dots & \dots \end{pmatrix}^{30} \cdot \begin{pmatrix} 0.9 \\ 0.1 \end{pmatrix} = ?$$

Equilibrium: $A \cdot \underline{v} = \underline{v}$ $\left\{ \begin{array}{l} \underline{v} \text{ eigenvector for } A \\ \text{with } \lambda = 1. \end{array} \right.$

$$(A - 1 \cdot I) \underline{v} = \underline{0} \quad \begin{pmatrix} -0.002 & 0.136 \\ 0.002 & -0.136 \end{pmatrix} \cdot \begin{pmatrix} v_1 \\ v_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$-0.002 v_1 + 0.136 v_2 = 0$$

$$v_1 = \frac{0.136}{0.002} v_2 = 68 v_2$$

$$\underline{v} = \begin{pmatrix} v_1 \\ v_2 \end{pmatrix} = \begin{pmatrix} 68 v_2 \\ v_2 \end{pmatrix} = v_2 \cdot \begin{pmatrix} 68 \\ 1 \end{pmatrix}$$

$$v_2 = \frac{1}{69} : \underline{v} = \underline{\underline{\begin{pmatrix} 68/69 \\ 1/69 \end{pmatrix}}}$$

Fact: With this kind of set-up, there is a unique equilibrium state \underline{v} , and $\lim_{T \rightarrow \infty} A^T \underline{v}_0 = \underline{v}$ for any \underline{v}_0 .

②

Diagonalization

A : $n \times n$ -matrix

Defn: A is diagonalizable if there is a diagonal matrix D and an invertible matrix P such that

$$P^{-1} \cdot A \cdot P = D$$

Motivation:

If $P^{-1}AP = D$, then

$$\boxed{A = PDP^{-1}}$$

This means:

$$A^n = \underbrace{A \cdot A \cdot A \cdot \dots \cdot A}_n = \underbrace{(PDP^{-1})(PDP^{-1})(PDP^{-1}) \dots (PDP^{-1})}_n$$

$$\boxed{A^n = PD^nP^{-1}}$$

$$D = \begin{pmatrix} 0.1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & -0.2 \end{pmatrix}$$

$$D^n = \begin{pmatrix} 0.1^n & 0 & 0 \\ 0 & 1^n & 0 \\ 0 & 0 & (-0.2)^n \end{pmatrix}$$

(easy to compute D^n
when D is diagonal,
difficult to compute A^n)

How to find a diagonalization of A : (if it is possible)

* find all eigenvalues of A

If there are n eigenvalues $\lambda_1, \lambda_2, \dots, \lambda_n$,
then we put

$$D = \begin{pmatrix} \lambda_1 & 0 & 0 & \dots & 0 \\ 0 & \lambda_2 & 0 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & \lambda_n \end{pmatrix}$$

* find all eigenvectors of A for $\lambda_1, \dots, \lambda_n$:

If there are n linearly independent eigenvectors,
then we put

$$P = \left(\begin{array}{c} \underline{v_1} \\ \underline{v_2} \\ \vdots \\ \underline{v_n} \end{array} \right)$$

and A is diagonalizable. If not, A is not diagonalizable.

Fact: If $\lambda = \lambda_i$ has m_i degrees of freedom, i.e. if

$(A - \lambda_i I)\underline{x} = 0$ has m_i degrees of freedom

then $\lambda = \lambda_i$ will contribute with m_i linearly independent eigenvectors.

Ex 1

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

$$\begin{aligned} \lambda_1 &= 1 \\ \lambda_2 &= 3 \\ \lambda_3 &= 1 \end{aligned}$$

$$\begin{aligned} \lambda &= 1 && (\text{mult. } 2) \\ \lambda &= 3 && (\text{mult. } 1) \end{aligned}$$

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

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

Yes, A is diagonalizable.

$$P^{-1}AP = D$$

\Downarrow

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

\Downarrow

$$A^n = PD^nP^{-1}$$

$$A^n = \begin{pmatrix} 0 & -1 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 1 \end{pmatrix} \cdot \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 3^n \end{pmatrix} \cdot \begin{pmatrix} 0 & -1 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 1 \end{pmatrix}^{-1}$$

$\lambda = 3$:

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

$$\underline{x} = \begin{pmatrix} z \\ 0 \\ z \end{pmatrix} = z \cdot \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} = \underline{v_3}$$

$\lambda = 1$:

$$\begin{pmatrix} 1 & 0 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} \underline{\lambda = 0}$$

$$\begin{aligned} \underline{x} = \begin{pmatrix} z \\ y \\ z \end{pmatrix} &= \begin{pmatrix} 0 \\ y \\ 0 \end{pmatrix} + \begin{pmatrix} -z \\ 0 \\ z \end{pmatrix} \\ &= y \cdot \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} + z \cdot \begin{pmatrix} -1 \\ 0 \\ 1 \end{pmatrix} \\ &= \underline{v_1} + \underline{v_2} \end{aligned}$$

Some facts:

- ① If a linear system $A \cdot \underline{x} = 0$ has m degrees of freedom, then there are m free variables s_1, s_2, \dots, s_m , and the solutions can be written in standard form as

$$\underline{x} = s_1 \underline{v}_1 + s_2 \underline{v}_2 + \dots + s_m \underline{v}_m$$

use Gaussian elimination to find solutions.

Then $\{\underline{v}_1, \underline{v}_2, \dots, \underline{v}_m\}$ are linearly independent vectors

- ② When you combine eigenvectors corresponding to different eigenvalues, you get linearly independent vectors.

- ③ A $n \times n$ -matrix:

A diagonalizable \iff A has n linearly independent eigenvectors



- i) A has n (real) eigenvalues, counted with multiplicity
and
ii) For each eigenvalue λ , the multiplicity of λ equals the number of degrees of freedom of $(A - \lambda I)\underline{x} = \underline{0}$.

- ④ If A is symmetric, then A is diagonalizable.