

LECTURE 32

Time Invariant ("Autonomous") Linear Systems.

I. Solution via Taylor Series

II. Solution via Eigenvalues & Eigenvectors

A) Algebraic Analysis

B) Comments

C) Geometric Analysis

D) Recommendation

III. Diagonal Form of a Matrix } ^{next} Lecture

TIME INVARIANT (AUTONOMOUS)

LINEAR SYSTEMS

Consider the system of differential equations with constant coefficients

$$\frac{d\vec{u}(t)}{dt} = A\vec{u}(t) \quad A: V = \mathbb{R}^n \rightarrow V = \mathbb{R}^n$$

Find $\vec{u}(t)$ subject to the initial condition $\vec{u}(0) = \vec{u}_0$

Comment:

Here we develop the mathematical theory for a linear system represented by a constant matrix A .

There are two mathematical ways of solving this dynamical system problem namely via

I. Taylor series method

II. Eigen vector method

I. Taylor Series Method

This method can be used for any matrix

A as long as it is a constant matrix. Indeed one has

$$\frac{d^1 u}{dt^1} = A u(t)$$

$$\frac{d^2 u}{dt^2} = A \frac{du}{dt} = A^2 u(t)$$

$$\frac{d^3 u}{dt^3} = A^3 u$$

The Taylor series expansion around $t=0$

yields

$$u(t) = u(0) + \frac{du}{dt}\bigg|_0 t + \frac{d^2 u}{dt^2}\bigg|_0 \frac{t^2}{2!} + \dots$$

$$= \underbrace{\left[I + At + \frac{1}{2!} A^2 t^2 + \dots \right]}_{\equiv e^{At}} u(0)$$

Thus one has e^{At} whenever the series converges

$$\vec{u}(t) = e^{At} \vec{u}_0$$

and

$$\vec{u}(t_1, t_2) = e^{At_1} e^{At_2} \vec{u}_0 = e^{A(t_1+t_2)} \vec{u}_0$$

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One says that e^{At} is the time evolution operator of the system,
Aside from the fact that A is said to be the

generator of time translations of
the system, there is not too much one
can say about it without examining
the structure of the $n \times n$ matrix A .

This deficiency is remedied by the

II. Eigenvector Method

The virtue of this method is that it
finds and uses the eigenvalues and eigenvectors
of A and thereby reveals the internal
mathematical and physical structure
of the linear system. This means
one will be able to predict the evolution
 $\vec{u}(t)$ for all possible initial value
 $\vec{u}(t=0) = \vec{u}_0$.

A) Algebraic Analysis

The motivation for the eigenvector
method comes from the familiar
one-dimensional equation

$$\frac{du}{dt} = Au,$$

Its solution is

$$u(t) = f(t)c$$

where c is an arbitrary constant
and

$$f(t) = e^{At}$$

Q: How does one apply this method if
 A is $n \times n$?

A: We are looking for an n -dimensional
vector \vec{u} . This suggests that we
look for solutions having the form

$$\vec{u}(t) = f(t)\vec{x}$$

where \vec{x} is a constant vector. Physically
this means that all components of

$\vec{u}(t)$ have the same time dependence.

Inserting this type of a solution into the

equation

$$\frac{d\vec{u}}{dt} = A\vec{u}$$

yields

$$A\vec{x} = \frac{d\vec{x}}{dt}$$

or

$$A\vec{x} = \frac{1}{f} \frac{d\vec{x}}{dt}$$

independent
of time

This equation implies:

$$(i) \quad \frac{df}{dt} = \text{constant} = \lambda \Rightarrow f = e^{\lambda t}$$

$$(ii) \quad (A - \lambda I)\vec{x} = 0$$

One must find values λ for which
 $N(A - \lambda I) \neq \{0\}$

For each such value of λ the solution is

$$\vec{u}(t) = e^{\lambda t} \vec{x} \quad \text{where } \vec{x} \in N(A - \lambda I)$$

B) Comments

(i) $\vec{u}(t) = e^{\lambda t} \vec{x}$ is called an eigen state.
(or an eigen mode if λ is complex)

Comment (continued)

(i') The linearity of the system implies that the

$$\text{(general solution)} = c_1 e^{\lambda_1 t} \vec{x}_1 + c_2 e^{\lambda_2 t} \vec{x}_2 + \dots + c_n e^{\lambda_n t} \vec{x}_n$$

whenever there are n distinct eigenvalues

(ii) We obtain such a solution only for a time invariant system

(iv) For λ to be an eigenvalue of A the following conditions are equivalent

$$(1) \quad Ax = \lambda x \quad \text{has a nontrivial sol}^n$$

$$(2) \quad A - \lambda I \quad \text{is singular (columns form a linearly dependent set)}$$

$$(3) \quad A - \lambda I \quad \text{has linearly dependent rows}$$

$$(4) \quad \det(A - \lambda I) = 0$$

Example

$$\frac{d\vec{u}}{dt} = \begin{bmatrix} 4 & -5 \\ 2 & -3 \end{bmatrix} \vec{u} \text{ with } \vec{u}(0) = \begin{bmatrix} 8 \\ 5 \end{bmatrix}$$

Algebraic analysis

Step 1. $Ax = \lambda x$
 $\det(A - \lambda I) = 0 \Rightarrow \lambda^2 - \lambda - 2 = 0$
 $(\lambda - 2)(\lambda + 1) = 0$

$\lambda_1 = -1$; compute $\vec{x}_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$

$\lambda_2 = 2$; compute $\vec{x}_2 = \begin{bmatrix} 5 \\ 2 \end{bmatrix}$

Step 2. Write down the general solution

$$\vec{u} = c_1 e^{\lambda_1 t} \begin{bmatrix} 1 \\ 1 \end{bmatrix} + c_2 e^{\lambda_2 t} \begin{bmatrix} 5 \\ 2 \end{bmatrix}$$

Step 3. Write down the solution which satisfies the initial conditions

$$\vec{u}(0) = \begin{bmatrix} 8 \\ 5 \end{bmatrix} \Rightarrow \vec{u}(t) = 3e^{\lambda_1 t} \begin{bmatrix} 1 \\ 1 \end{bmatrix} + 1e^{\lambda_2 t} \begin{bmatrix} 5 \\ 2 \end{bmatrix}$$
$$= \begin{bmatrix} 3e^{-t} + 5e^{2t} \\ 3e^{-t} + 2e^{2t} \end{bmatrix}$$

c) Geometrical Analysis

The geometrical method applies to linear as well as to non-linear systems.

Consider

$$\left. \begin{aligned} \frac{dv}{dt} = f(v, w) &= 4v - 5w \\ \frac{dw}{dt} = g(v, w) &= 2v - 3w \end{aligned} \right\} (*)$$

Here $(v(t), w(t)) =$ position in $V = \mathbb{R}^2$ as a fn of time.

$\left(\frac{dv}{dt}, \frac{dw}{dt} \right) =$ velocity at point $(v(t), w(t))$

$\begin{matrix} \nearrow \\ (v, w) \end{matrix}$ = tangent to curve

$$\frac{dw}{dv} = \frac{dw/dt}{dv/dt} = \frac{g(v, w)}{f(v, w)} = \text{slope of tangent at } (v, w)$$

Problem

a) Sketch the phase portrait of Eq. (*) for which one has

$$\frac{dw}{dv} = \frac{2v - 3w}{4v - 5w}$$

b) Exhibit the tangent at strategic pts.

Solution:

$$\frac{d}{dt} \begin{bmatrix} v \\ w \end{bmatrix} = \begin{bmatrix} 2 & -3 \\ 4 & -5 \end{bmatrix} \begin{bmatrix} v \\ w \end{bmatrix}$$

$$A: \lambda_1 = -1 \Rightarrow \begin{bmatrix} v \\ w \end{bmatrix} \propto \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \lambda_2 = 2 \Rightarrow \begin{bmatrix} v \\ w \end{bmatrix} \propto \begin{bmatrix} 5 \\ 2 \end{bmatrix}$$

What is the phase space diagram in the $v-w$ plane?

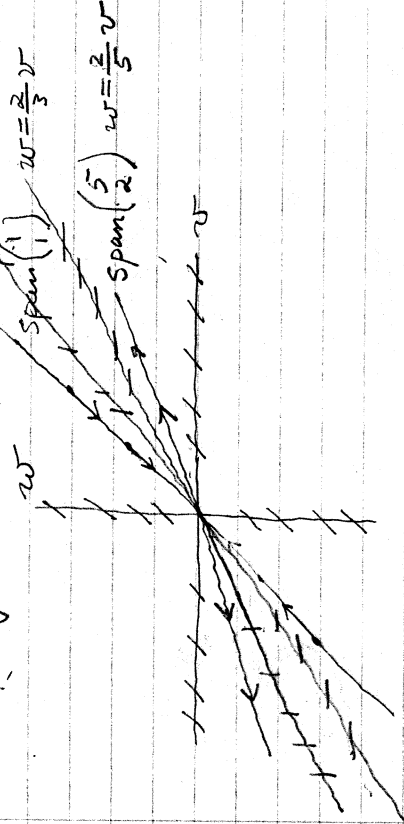
$$\frac{w}{v} = \frac{2}{5}$$

Let $\begin{bmatrix} v \\ w \end{bmatrix} \in \mathcal{N}(A - \lambda I)$

$$\frac{d}{dt} \begin{bmatrix} v \\ w \end{bmatrix} = A \begin{bmatrix} v \\ w \end{bmatrix} = \lambda \begin{bmatrix} v \\ w \end{bmatrix}$$

velocity

position



$$\frac{dw}{dv} = \frac{2v - 3w}{4v - 5w}$$

$$w=0 \Rightarrow \frac{dw}{dv} = \frac{1}{2}$$

$$v=0 \Rightarrow \frac{dw}{dv} = \frac{3}{5}$$

$$w = \frac{2}{3}v \Rightarrow \frac{dw}{dv} = 0$$

$$w = \frac{4}{5}v \Rightarrow \frac{dw}{dv} = \infty$$

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Slope:

$$\frac{dw}{dv} = 1 \text{ (eigendirection)}$$

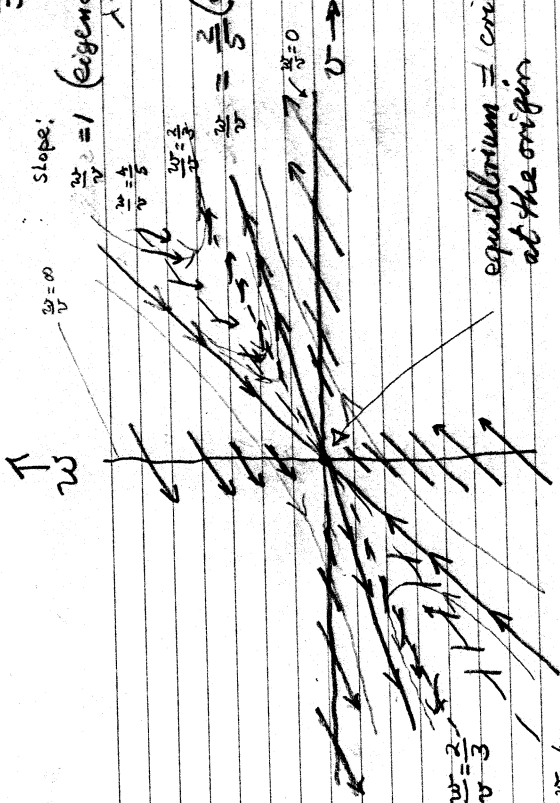
$$\frac{dw}{dv} = \frac{4}{3}$$

$$\frac{dw}{dv} = \frac{2}{5}$$

$$\frac{dw}{dv} = \frac{2}{5} \text{ (eigendirection)}$$

$$\lambda = -1$$

$$\lambda = 2$$



equilibrium = critical point at the origin

$$\frac{dv}{dt} = 0: \frac{w}{v} = \frac{2}{3}$$

$$\frac{dw}{dt} = 0: \frac{w}{v} = \frac{4}{5}$$

$$v = \frac{5}{4}w \quad \frac{dw}{dt} = (2 - \frac{5}{4} - 3)w$$

$$= (\frac{5}{2} - 3)w = -\frac{1}{2}w$$

D) Recommendation!

Use "quiver" command in Matlab

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III. DIAGONAL FORM OF A MATRIX

Read Strang 5.2

The essential properties of a $n \times n$ matrix with a set of n linearly independent eigen vectors are expressed in terms of its eigenvalues and the transition matrix formed from these eigen vectors. This is expressed by the following

Proposition

Let $A: \mathbb{R}^n \rightarrow \mathbb{R}^n$ be an $n \times n$ matrix. Let A have a set of n lin. indep. eigen vectors. Let S be the transition matrix formed from these eigen values.

Then

$$S^{-1}AS = \begin{bmatrix} \lambda_1 & & 0 \\ & \lambda_2 & \\ 0 & & \lambda_n \end{bmatrix}$$

One says that A can be diagonalized

proof:

Consider the equations

$$A\vec{x}_i = \lambda_i \vec{x}_i \quad i=1, 2, \dots, n$$

satisfied by the given eigen vectors.

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One has

$$A \begin{bmatrix} \vec{x}_1 \\ \vdots \\ \vec{x}_m \end{bmatrix} = \begin{bmatrix} \lambda_1 \vec{x}_1 & \dots & \lambda_m \vec{x}_m \end{bmatrix}$$

$$= \begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_m \end{bmatrix} \begin{bmatrix} \vec{x}_1 \\ \vdots \\ \vec{x}_m \end{bmatrix}$$

or $AS = SA$ where S is invertible (why?)

Thus $S^{-1}AS = \Lambda$ or $A = SAS^{-1}$

A is diagonalizable by means of a similarity transformation.

Comments:

1. It is easy to compute powers of a diagonalizable matrix

$$A^m = (S\Lambda S^{-1}) \dots (S\Lambda S^{-1})$$

$$= S \Lambda^m S^{-1}$$

2. One can compute matrix power series. Consider a function defined by a convergent power series

$$g(x) = a_0 + a_1 x + a_2 x^2 + \dots; \text{ e.g. } e^x = 1 + x + \frac{x^2}{2} + \dots$$

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Then $g(A)$ is defined and convergent in an appropriate sense:

$$\begin{aligned} e^{At} &= I + At + A^2 \frac{t^2}{2} + \dots \\ &= S I S^{-1} + S A S^{-1} t + S A^2 S^{-1} \frac{t^2}{2!} + \dots \\ &= S \left(I + \lambda t + \frac{\lambda^2 t^2}{2!} + \dots \right) S^{-1} \\ &= S \begin{bmatrix} e^{\lambda_1 t} & & 0 \\ & \ddots & \\ 0 & & e^{\lambda_n t} \end{bmatrix} S^{-1} \end{aligned}$$

$$= S e^{\lambda t} S^{-1}$$

Thus

$$g(A) = S g(\lambda) S^{-1}$$

3. Consider the secular (a.k.a, characteristic) Polynomial

$$\begin{aligned} \det(A - \lambda I) &= a_0 + a_1(-\lambda) + \dots + a_n(-\lambda)^n \\ &= (\lambda_1 - \lambda)(\lambda_2 - \lambda) \dots (\lambda_n - \lambda) = f(\lambda) \end{aligned}$$

The secular pol' f in the powers of λ vanishes!

$$\begin{aligned} f(A) &= S^{-1} f(\lambda) S = \\ &= S^{-1} \begin{bmatrix} (\lambda_1 - A)(\lambda_2 - A) \dots (\lambda_n - A) & & 0 \\ & \ddots & \\ 0 & & (\lambda_1 - A) \dots (\lambda_n - A) \end{bmatrix} S = S^{-1} \begin{bmatrix} 0 & & 0 \\ & \ddots & \\ 0 & & 0 \end{bmatrix} S \\ &= \begin{bmatrix} 0 & & 0 \\ & \ddots & \\ 0 & & 0 \end{bmatrix} = \text{[Zero matrix]} \end{aligned}$$

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Thus

$$f(A) = a_0 + a_1(A) + a_2(A)^2 + \dots + a_n(A)^n = \text{zero matrix}$$

This illustrates the Cayley - Hamilton theorem:

A matrix satisfies its characteristic equation

Here the coefficients

$$a_0 = f(0) = \lambda_1 \lambda_2 \dots \lambda_n = \det A$$

$$a_{n-1} = \lambda_1 + \dots + \lambda_n = \sum_{i=1}^n A_{ii} = \text{trace } A$$

depend only on the eigenvalues of A .