

# Lec 5

Story of

$$\dot{x}(t) = A(t)x(t)$$

①

On the story of

$$\dot{X}(t) = A(t) X(t)$$

Let us assume that  $A$  is a square  $n \times n$  matrix whose elements are continuous fns of time on the interval  $0 \leq t \leq T$ .

We define a sequence of matrices  $M_k$  as follows:

$$M_0 = I$$

$$M_1 = I + \int_0^t A(\sigma) M_0(\sigma) d\sigma$$

...

$$M_k = I + \int_0^t A(\sigma) M_{k-1}(\sigma) d\sigma.$$

Theorem:

The sequence of matrices  $M_0, M_1, \dots$  converges uniformly on the interval  $[0, T]$ . If we denote the limit function by  $\phi(t)$  then for  $0 \leq t \leq T$ , we have

$$\dot{\phi}(t) = A(t)\phi(t)$$

$$\& \phi(0) = I.$$

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One can rewrite  $M_k$  as follows.

$$M_0 = I$$

$$M_1 = M_0 + \int_0^t A(\sigma_1) d\sigma_1.$$

$$M_2 = M_1 + \int_0^t A(\sigma_1) \left[ \int_0^{\sigma_1} A(\sigma_2) d\sigma_2 \right] d\sigma_1.$$

.....

$$M_k = M_{k-1} + \int_0^t A(\sigma_1) \int_0^{\sigma_1} A(\sigma_2) \int_0^{\sigma_2} A(\sigma_3) \dots$$

$$\dots \int_0^{\sigma_{k-1}} A(\sigma_k) d\sigma_k d\sigma_{k-1} \dots d\sigma_1$$

$P_k(t)$

The series

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$$I + P_1(t) + P_2(t) + \dots$$

is called the Peano-Baker series

The theorem on page 2 says that the P-B series converges uniformly in the interval  $[0, T]$ . We define

$$\phi(t) = \sum_{j=0}^{\infty} P_j(t) \quad P_0 = I.$$

Lemma:

If  $A(t)$  is a scalar continuous f<sup>n</sup> of  $t$ , i.e. if we write  $A(t) = a(t)$ , then

$$\phi(t) = \exp \left[ \int_0^t a(\sigma) d\sigma \right].$$

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Proof:

$$P_0 = 1$$

$$P_1 = \int_0^t a(\sigma) d\sigma$$

$$P_2 = \int_0^t a(\sigma_1) \int_0^{\sigma_1} a(\sigma_2) d\sigma_2 d\sigma_1.$$

$$= \left[ \int_0^{\sigma_1} a(\sigma_2) d\sigma_2 \right] \Big|_0^t P_1 - \underbrace{\int_0^t a(\sigma_1) \int_0^{\sigma_1} a(\sigma_2) d\sigma_2 d\sigma_1}_{P_2}$$

$$\Rightarrow P_2 = P_1^2 - P_2$$

$$\Rightarrow P_2 = \frac{1}{2} P_1^2 = \frac{1}{2} \left[ \int_0^t a(\sigma) d\sigma \right]^2$$

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Assume that

$$P_k(t) = \frac{1}{k!} \left[ \int_0^t a(\sigma) d\sigma \right]^k$$

We write

$$P_{k+1}(t) = \int_0^t a(\sigma) P_k(\sigma) d\sigma$$

$$= P_k(\sigma) \Big|_0^t P_1(t)$$

$$= \int_0^t \frac{1}{(k-1)!} \left[ \int_0^{\sigma_2} a(\sigma) d\sigma \right]^{k-1} a(\sigma_2) P_1(\sigma_2) d\sigma_2$$

$$= P_k(t) P_1(t) - \int_0^t \frac{1}{(k-1)!} a(\sigma_2) \left[ \int_0^{\sigma_2} a(\sigma) d\sigma \right]^k$$

$$= P_k(t) P_1(t) - k \int_0^t a(\sigma_2) P_k(\sigma_2) d\sigma_2$$

$$= P_k(t) P_1(t) - k P_{k+1}(t).$$

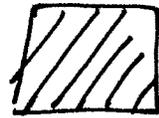
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Hence

$$(k+1) P_{k+1}(t) = P_k(t) P_1(t)$$

$$\Rightarrow P_{k+1}(t) = \frac{1}{k+1} \cdot \frac{1}{k!} [P_1(t)]^k P_1(t)$$

$$= \frac{1}{(k+1)!} [P_1(t)]^{k+1}.$$



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## Proof of the theorem on page 2

In order to prove that the sequence  $M_k(t)$  converges, we need to show that the scalar sequence consisting of  $(i, j)^{th}$  element converges for  $i=1, \dots, n; j=1, 2, \dots, n$ . etc.

Recall that a series of continuous scalar functions  $x_1(t) + x_2(t) + \dots$  defined on  $[0, T]$  converges absolutely and uniformly on the interval  $[0, T]$  if  $\exists$  sequence of positive constants  $c_i$ :  $\forall t \in [0, T]$  we have  $|x_i(t)| \leq c_i$  and the series  $c_1 + c_2 + \dots$  converges.

"This is called Weierstrass M-test"

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Define  $\eta(t)$  as the maximum absolute value of any entry in  $A(t)$ . i.e

$$\eta(t) = \max_{i,j} |a_{ij}(t)|$$

$$\text{Let } \gamma(t) = \int_0^t \eta(\sigma) d\sigma.$$

Let us denote

$$E_{ij}(Q)$$

to be the  $i,j^{\text{th}}$  element of the matrix  $Q$ .

Let  $P_k(t)$  be the  $k^{\text{th}}$  term of the Peano-Baker series (page 4). We have

$$E_{ij}(P_k(t)) = E_{ij} \left( \int_0^t \int_0^{\sigma_1} \dots \int_0^{\sigma_{k-1}} A(\sigma_1) \dots A(\sigma_k) d\sigma_1 \dots d\sigma_k \right)$$

If  $Q_1$  &  $Q_2$  are two  $n \times n$  matrices, we have

$$|E_{ij}(Q_1, Q_2)| \leq n \max_{ij} |E_{ij}(Q_1)|$$

$$\cdot \max_{ij} |E_{ij}(Q_2)|.$$

It would follow that

$$E_{ij}(P_k(t)) \leq \int_0^t \int_0^{\sigma_1} \dots \int_0^{\sigma_{k-1}} n^{k-1} \eta(\sigma_1) \eta(\sigma_2) \dots \eta(\sigma_k) d\sigma_1 \dots d\sigma_k.$$

$$= n^{k-1} \frac{\gamma^k(t)}{k!}$$

If we define

$$C_k(t) = n^{k-1} \frac{r^k(t)}{k!}$$

it follows that

$$E_{ij}(P_k(t)) \leq C_k(t)$$

Moreover

$$\begin{aligned} & C_0 + C_1 + C_2 + \dots \\ &= 1 + r(t) + \frac{n r^2(t)}{2!} + \dots \\ &= 1 + \frac{1}{n} \left( n r + \frac{n^2 r^2}{2!} + \dots \right) \end{aligned}$$

$$= 1 + \frac{1}{n} (e^{nr} - 1)$$

Hence  $E_{ij}(P_k(t))$  converges for every  $i, j$ .

To show that

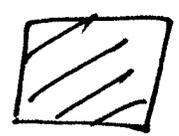
$$\dot{\phi}(t) = A(t) \phi(t)$$

recall that

$$\begin{aligned} \phi(t) = & I + \int_0^t A(\sigma_1) d\sigma_1 \\ & + \int_0^t \int_0^{\sigma_1} A(\sigma_1) A(\sigma_2) d\sigma_1 d\sigma_2 . \\ & + \dots \end{aligned}$$

It follows that

$$\begin{aligned} \dot{\phi}(t) = & 0 + A(t) + A(t) \int_0^t A(\sigma_1) d\sigma_1 . \\ & + A(t) \int_0^t \int_0^{\sigma_1} A(\sigma_1) A(\sigma_2) d\sigma_1 d\sigma_2 \\ & \dots \\ = & A(t) \phi(t) . \end{aligned}$$



Consider the o.d.e

$$\dot{x} = A(t)x$$

$$x(0) = x_0.$$

We have

$$x(t) = \phi(t)x_0.$$

To see this note that

$$\dot{x}(t) = \dot{\phi}(t)x_0 = A(t)\phi(t)x_0 = A(t)x(t).$$

Moreover

$$x(0) = \phi(0)x_0 = I x_0 = x_0.$$

The matrix  $\phi(t)$  is called the transition matrix.

Corollary:

If  $A$  is a real constant  $n \times n$  matrix

then 
$$\phi(t) = I + At + \frac{A^2 t^2}{2!} + \frac{A^3 t^3}{3!} + \dots$$

$$\triangleq e^{At}$$

Proof:

$$P_1 = \int_0^t A d\sigma = A \int_0^t d\sigma = At$$

$$P_2 = \int_0^t \int_0^{\sigma_1} A^2 d\sigma_2 d\sigma_1$$

$$= A^2 t^2 / 2!$$

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Example

consider the 2<sup>nd</sup> order eq<sup>n</sup>

$$\begin{pmatrix} \dot{x}_1(t) \\ \dot{x}_2(t) \end{pmatrix} = \begin{pmatrix} \sigma(t) - \omega(t) & \\ +\omega(t) & \sigma(t) \end{pmatrix} \begin{pmatrix} x_1(t) \\ x_2(t) \end{pmatrix}$$

Calculate the transition matrix.

Writing  $z(t) = x_1(t) + i x_2(t)$

$$a(t) = \sigma(t) + i\omega(t)$$

we have

$$\dot{z}(t) = \dot{x}_1(t) + i\dot{x}_2(t)$$

$$= (\sigma x_1 - \omega x_2) + i(\omega x_1 + \sigma x_2)$$

$$= (\sigma + i\omega) x_1 + i(\sigma + i\omega) x_2$$

$$= a(t) z(t).$$

∴ ∴  $a(t)$  is scalar, it follows that

$$z(t) = e^{\int_0^t a(\tau) d\tau} z(0).$$

Hence  $\int_0^t [\sigma(\tau) + i\omega(\tau)] d\tau$ .

$$x_1(t) + ix_2(t) = e^{\int_0^t \sigma(\tau) d\tau} [x_1(0) + ix_2(0)].$$

$$= e^{\int_0^t \sigma(\tau) d\tau} \left[ \cos \int_0^t \omega(\tau) d\tau + i \sin \int_0^t \omega(\tau) d\tau \right] (x_1(0) + ix_2(0)).$$

We write

$$\begin{pmatrix} x_1(t) \\ x_2(t) \end{pmatrix} = e^{\int_0^t \sigma(\tau) d\tau} \begin{pmatrix} \cos \int_0^t \omega(\tau) d\tau & -\sin \int_0^t \omega(\tau) d\tau \\ \sin \int_0^t \omega(\tau) d\tau & \cos \int_0^t \omega(\tau) d\tau \end{pmatrix} \begin{pmatrix} x_1(0) \\ x_2(0) \end{pmatrix}$$

The transition matrix

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$$\Phi(t) =$$

$$e^{\int_0^t \sigma(\tau) d\tau} \begin{pmatrix} \cos \int_0^t \omega(\tau) d\tau & -\sin \int_0^t \omega(\tau) d\tau \\ \sin \int_0^t \omega(\tau) d\tau & \cos \int_0^t \omega(\tau) d\tau \end{pmatrix}$$

— x —

Remark: If  $\sigma$  and  $\omega$  are constants we have

$$\Phi(t) = \begin{pmatrix} e^{\sigma t} \cos \omega t & -e^{\sigma t} \sin \omega t \\ e^{\sigma t} \sin \omega t & e^{\sigma t} \cos \omega t \end{pmatrix}$$

# Example

Consider a 2<sup>nd</sup> order o.d.e

$$\ddot{y} + 3\dot{y} + \omega(t)^2 y = 0$$

where we assume that

$$\dot{\omega} = -3\omega \quad \omega(0) = \omega_0$$

$$y(0) = y_0$$

$$\dot{y}(0) = y_1$$

Problem: Calculate  $y(t)$ .

Define  $x_2(t) = y(t)$

$$x_1(t) = \frac{\dot{y}(t)}{\omega(t)}$$

We have

$$\dot{x}_2 = \dot{y} = \omega x_1$$

$$\dot{x}_1 = \frac{\omega \ddot{y} - \dot{y} \dot{\omega}}{\omega^2} = \frac{\ddot{y}}{\omega} + \frac{\dot{y}}{\omega} 3$$

$$= \frac{-3\dot{y} - \omega^2 y}{\omega} + 3 \frac{\dot{y}}{\omega} = -\omega y = -\omega x_2$$

∴ we have

$$\begin{pmatrix} \dot{x}_1(t) \\ \dot{x}_2(t) \end{pmatrix} = \begin{pmatrix} 0 & \omega(t) \\ -\omega(t) & 0 \end{pmatrix} \begin{pmatrix} x_1(t) \\ x_2(t) \end{pmatrix}$$

$$\omega(t) = e^{-3t} \omega_0$$

Rotation with an exponentially decaying angular velocity.

$$\begin{pmatrix} x_1 \\ x_2 \end{pmatrix}(t) = \begin{bmatrix} \cos \int_0^t e^{-3\tau} \omega_0 d\tau & -\sin \int_0^t e^{-3\tau} \omega_0 d\tau \\ \sin \int_0^t e^{-3\tau} \omega_0 d\tau & \cos \int_0^t e^{-3\tau} \omega_0 d\tau \end{bmatrix} \begin{pmatrix} x_1(0) \\ x_2(0) \end{pmatrix}$$

$$\int_0^t e^{-3\tau} d\tau = \left. \frac{e^{-3\tau}}{-3} \right|_0^t = -\frac{1}{3} (e^{-3t} - 1) = \frac{1}{3} (1 - e^{-3t}).$$

$$\begin{pmatrix} x_1 \\ x_2 \end{pmatrix}(t) =$$

$$\begin{pmatrix} \cos\left[\frac{\omega_0}{3}(1-e^{-3t})\right] & -\sin\left[\frac{\omega_0}{3}(1-e^{-3t})\right] \\ \sin\left[\frac{\omega_0}{3}(1-e^{-3t})\right] & \cos\left[\frac{\omega_0}{3}(1-e^{-3t})\right] \end{pmatrix}$$

$$\begin{pmatrix} x_1(0) \\ x_2(0) \end{pmatrix}$$

$$y(t) = x_2(t) =$$

$$x_1(0) \sin\left[\frac{\omega_0}{3}(1-e^{-3t})\right] +$$

$$x_2(0) \cos\left[\frac{\omega_0}{3}(1-e^{-3t})\right]$$

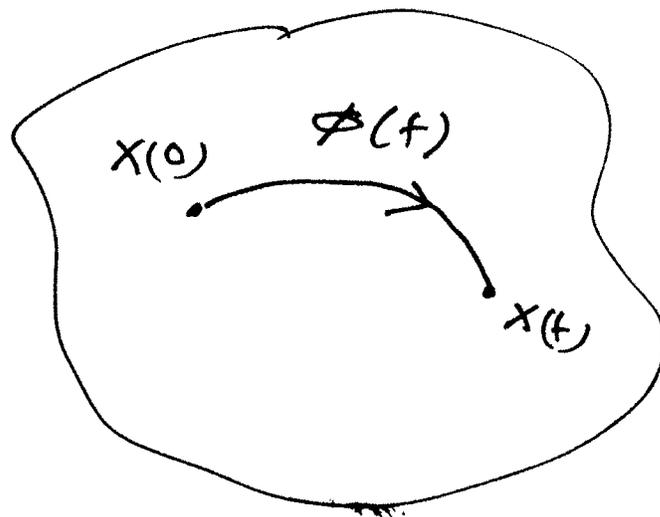
$$= \frac{y_1}{\omega_0} \sin\left[\frac{\omega_0}{3}(1-e^{-3t})\right] + y_0 \cos\left[\frac{\omega_0}{3}(1-e^{-3t})\right]$$



What does  $\phi(t)$  mean?!

Recall

$$x(t) = \phi(t) x(0)$$



$\phi(t)$  maps the vector  $x(0)$  at  $t=0$   
to the vector  $x(t)$  at  $t=t$

$$\phi(t): \mathbb{R}^n \rightarrow \mathbb{R}^n$$

$$x(0) \longmapsto x(t) = \phi(t) x(0).$$

Thus  $\phi(t)$  describes how the state evolves in time starting from  $t=0$ .

Often, we are interested in the transition of state from  $t = t_0$  instead of  $t = 0$ . It becomes necessary to define  $\Phi(t, t_0)$  as the state transition matrix which satisfies the matrix differential eq<sup>n</sup>

$$\dot{\Phi}(t, t_0) = A(t)\Phi(t, t_0)$$

where

$$\Phi(t_0, t_0) = \mathbf{I}.$$

Existence of  $\Phi(t, t_0)$  will follow from the Peano-Baker series where the lower limit of 0 on page 3 has to be replaced by  $t_0$ .

Here onwards, we shall denote by  $\Phi(t, t_0)$ , what we had denoted by  $\Phi(t)$  earlier.

Remark:

① What does  $\Phi(t_1, t_2)$  mean??

Solve matrix o.d.e.

$$\dot{M}(t) = A(t)M(t)$$

$$M(t_2) = I.$$

calculate  $M(t_1)$ . This would be  $\Phi(t_1, t_2)$ .

② Consider the vector o.d.e

$$\dot{x}(t) = A(t)x(t)$$

$$x(t_2) = x_0.$$

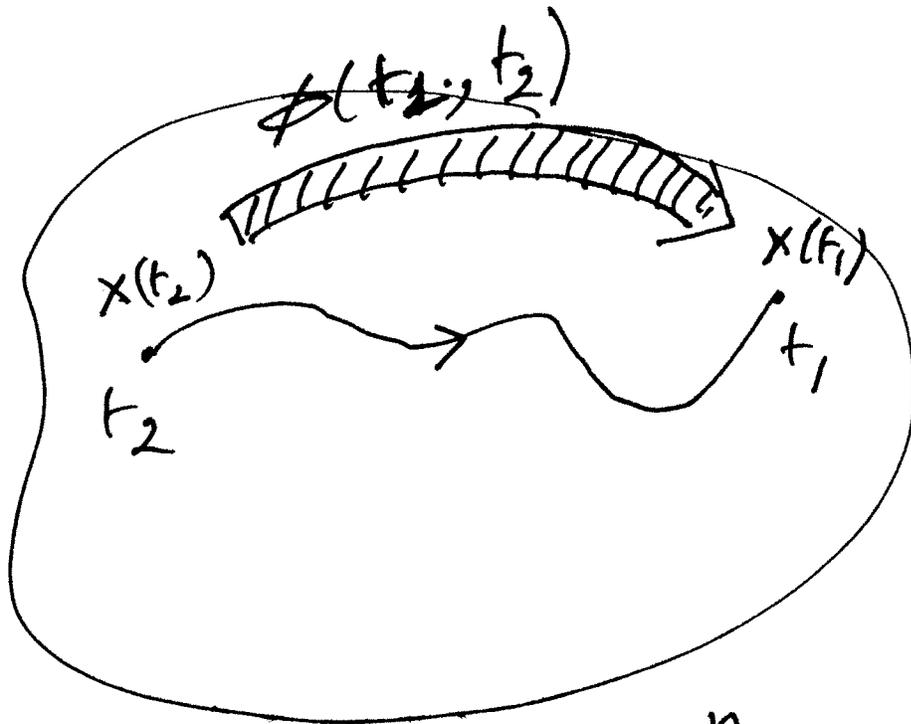
It would follow that

$$x(t) = \phi(t, t_2) x_0.$$

and

$$x(t_1) = \phi(t_1, t_2) x_0.$$

$\phi(t_1, t_2)$  maps the state from  $x_0$  at  $t = t_2$  to  $x(t_1)$  at  $t = t_1$ .



$$\phi(t_1, t_2) : \mathbb{R}^n \rightarrow \mathbb{R}^n.$$

$$x(t_2) \mapsto x(t_1) = \phi(t_1, t_2) x(t_2)$$

Q Theorem:

If  $t_1, t_2, t_3$  are 3 instances of time  
 where  $t_1 \leq t_2 \leq t_3$  we have

$$\phi(t_3, t_1) = \phi(t_3, t_2)\phi(t_2, t_1)$$

Theorem

If  $t_1, t_2, t_3$  are any 3 instances  
 of time (not necessarily ordered)

$$\phi(t_3, t_1) = \phi(t_3, t_2)\phi(t_2, t_1).$$

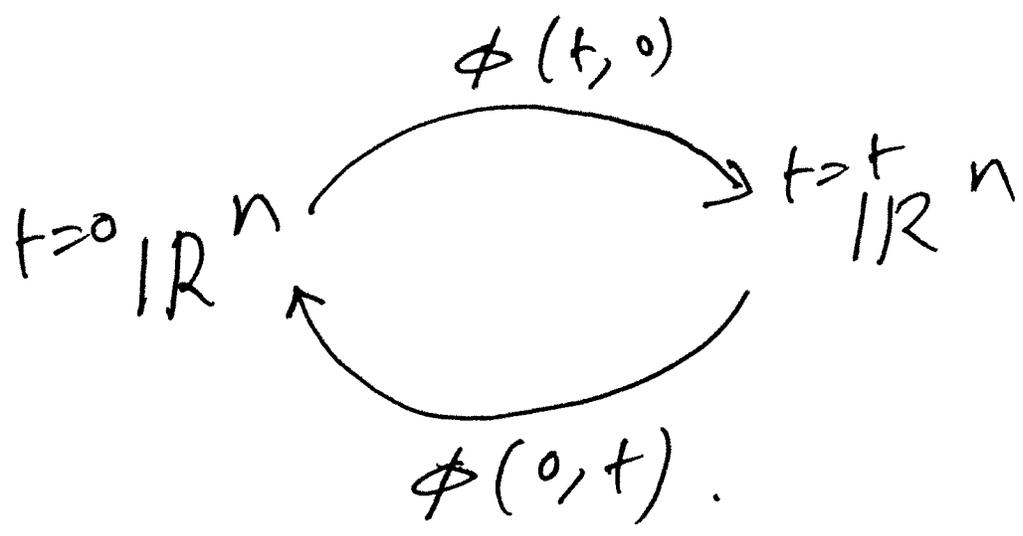
Cor: If  $t_3 = t_1$  we have

$$\begin{aligned} \phi(t_1, t_1) &= \phi(t_1, t_2)\phi(t_2, t_1) \\ &= \mathbb{I} \end{aligned}$$

Hence  $\phi(t_1, t_2) = \phi(t_2, t_1)^{-1}$ .

Choosing  $t_1 = 0, t_2 = t$  we have

$$\phi(0, t) = \phi(t, 0)^{-1}$$



$$\circ \circ \phi(0, t) \phi(t, 0) = I$$

it follows that

$$\frac{d}{dt} [\phi(0, t)] \phi(t, 0) + \phi(0, t) \frac{d}{dt} [\phi(t, 0)] = 0$$

$$\Rightarrow \frac{d}{dt} [\phi(0, t)] \cdot \phi(t, 0) + \phi(0, t) A \phi(t, 0) = 0$$

$$\Rightarrow \left[ \frac{d}{dt} \phi(0, t) + \phi(0, t) A \right]$$

$$\phi(t, 0) = 0 .$$

$$\forall \phi(t, 0) .$$

$$\Rightarrow \frac{d}{dt} \phi(0, t) = - \phi(0, t) A .$$

$$\Rightarrow \frac{d}{dt} \phi(t, 0)^{-1} = - \phi(t, 0)^{-1} A$$

Th<sup>m</sup> (Abel-Jacobi-Liouville)

If  $\phi(t, t_0)$  is the transition matrix for  $\dot{x}(t) = A(t)x(t)$  then  $\det[\phi(t, t_0)] = e^{\int_{t_0}^t \text{trace } A(\sigma) d\sigma}$

Hence  $\phi(t, t_0)$  is non-singular if  $\int_{t_0}^t \text{trace } A(\sigma) d\sigma$  is finite.

Proof: The theorem follows easily from this following fact:

Let  $M(t)$  be a square  $n \times n$  matrix then

$$\frac{d}{dt} [\det M(t)] = \text{trace}[C^T(t) \dot{M}(t)] \quad (*)$$

Where  $C^T(t)$  is the transpose of the cofactor matrix.

Replacing  $M(t)$  by  $\phi(t, t_0)$  we obtain

$$\begin{aligned} \frac{d}{dt} [\det \phi(t, t_0)] &= \\ & \text{trace} [C^T(t, t_0) A(t) \phi(t, t_0)] \\ &= \text{trace} [C^T(t, t_0) \phi(t, t_0) \cdot A(t)] \end{aligned}$$

However

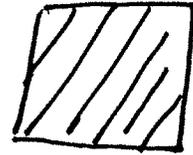
$$\begin{aligned} C^T(t, t_0) \phi(t, t_0) \\ = \det \phi(t, t_0) \cdot I \end{aligned}$$

Hence

$$\begin{aligned} \frac{d}{dt} [\det \phi(t, t_0)] &= \text{trace} [\det \phi(t, t_0) A(t)] \\ &= \det \phi(t, t_0) \text{trace } A(t) \end{aligned}$$

It follows that

$$\phi(t, t_0) = e^{\int_{t_0}^t \text{trace } A(\sigma) d\sigma}$$



Remark:

The determinant of an  $n \times n$  matrix can be interpreted as the volume in  $E^n$  contained in the parallelotope generated by the columns in the matrix

Thus Abel-Jacobi-Liouville's theorem gives us an expression for how this volume evolves in time. In particular if

$$\text{trace } A(t) = 0 \quad \forall t$$

we have

$$\det \Phi(t, t_0) = 1 \quad \forall t,$$

Hence the volume remains constant.

"Volume preserving flow".

$$\dot{x}(t) = A(t)x(t)$$

$$x(t_0) = x_0$$

$$\dot{z}(t) = B(t)z(t)$$

$$z(t_0) = z_0$$

Assume

$$z(t) = P(t)x(t)$$

$$\Rightarrow \dot{z}(t) = \dot{P}x + P\dot{x}$$

$$= \dot{P}x + PAx$$

$$= (\dot{P} + PA)x$$

$$= \underbrace{(\dot{P} + PA)P^{-1}}_B z$$

$$\dot{z} = Bz$$

$$z_0 = z(t_0) = P(t_0)x(t_0) = P(t_0)x_0.$$

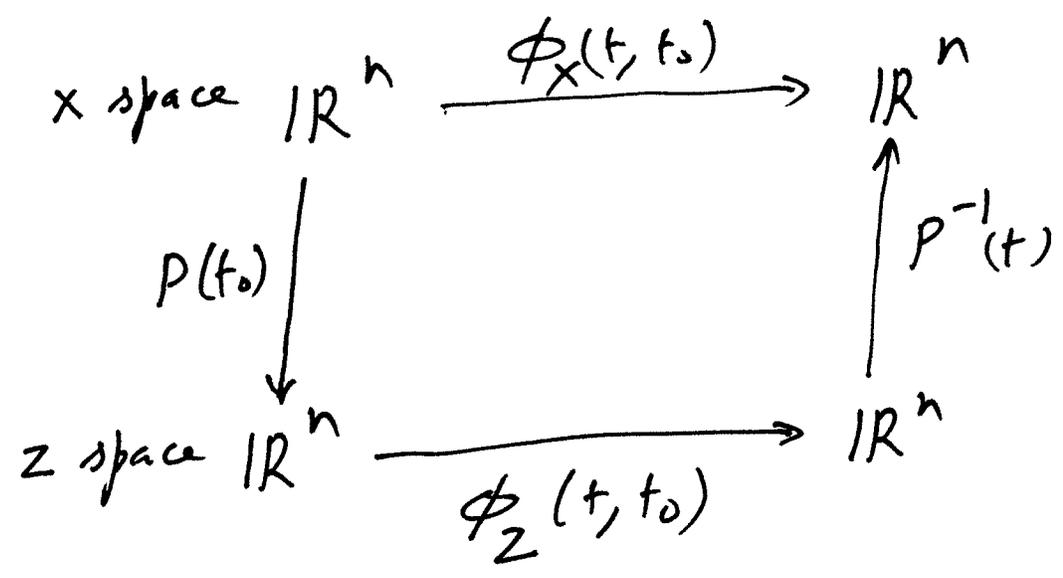
Define

$\Phi_x(t, t_0) \leftarrow$  transition for  $x$  system.

$\Phi_z(t, t_0) \leftarrow$  " " " " " "

Theorem

$$\Phi_X(t, t_0) = P^{-1}(t) \Phi_Z(t, t_0) P(t_0)$$



Proof:

$$\begin{aligned} \dot{\Phi}_X(t, t_0) &= \left[ \frac{d}{dt} P^{-1}(t) \right] \Phi_Z(t, t_0) P(t_0) \\ &\quad + P^{-1}(t) \frac{d}{dt} [\Phi_Z(t, t_0)] P(t_0) \\ &= -P^{-1}(t) \dot{P}^{-1} P^{-1} \Phi_Z(t, t_0) P(t_0) \\ &\quad + P^{-1}(t) B(t) \Phi_Z(t, t_0) P(t_0). \end{aligned}$$

$$= -P^{-1} \dot{P} P^{-1} \phi_2(t, t_0) P(t_0)$$

$$+ P^{-1} (\dot{P} + PA) P^{-1} \phi_3(t, t_0) P(t_0).$$

$$= P^{-1} P A P^{-1} \phi_3(t, t_0) P(t_0).$$

$$= A P^{-1} \phi_3(t, t_0) P(t_0)$$

$$= A \phi_x(t, t_0)$$

————— x —————

If we assume that  $B(t) = B$  a constant matrix we find a  $P(t)$ :

$$[\dot{P}(t) + P(t)A(t)] P^{-1}(t) = B$$

then  $\dot{x} = A(t)x(t)$

$$x(t_0) = x_0$$

can be transformed to a time invariant system

$$\dot{z}(t) = B z(t)$$

$$z(t_0) = P(t_0) x_0.$$

This transformation is not very useful since  $P(t)$  needs to be computed.

Remark:

We are often interested in placing additional constraints on  $P(t)$  as described below:

Def: A matrix  $P(t)$  is called a Lyapunov transformation if

- ①  $P(t), \dot{P}(t)$  are continuous and bounded functions on  $[t_0, \infty)$
- ②  $\exists$  a constant  $m!$   
 $0 < m < |\det(P(t))|.$

Linear Time Varying Dynamical eqns  
with periodic  $A(t)$ .

consider

$$\dot{x} = A(t)x(t) \quad \text{where } A(t+T) = A(t).$$

~~$$x(t_0) = x_0$$~~

Let  $\Phi_X(t, 0)$  be the transition matrix.

claim:

$\Phi_X(t+T, T)$  is also a transition matrix.

$$\text{ie } \frac{d}{dt} \Phi_X(t+T, T) = A(t) \Phi_X(t+T, T).$$

$$\begin{aligned} \text{Proof: } \frac{d}{dt} \Phi_X(t+T, T) &= A(t+T) \Phi_X(t+T, T) \\ &\quad \text{(From def)} \\ &= A(t) \Phi_X(t+T, T). \end{aligned}$$



$$\text{Hence } \Phi_X(t+T, T) = \Phi_X(t, 0) \quad \forall t.$$

Define  $Q = \Phi_{\underline{\lambda}}(T, 0)$ .

claim:

$$\Phi_{\underline{\lambda}}(t+T, 0) = \Phi_{\underline{\lambda}}(t, 0) Q.$$

Proof:

$$\Phi_{\underline{\lambda}}(T, 0) = Q \quad (\text{Def of } Q)$$

$$\begin{aligned} \Rightarrow \Phi_{\underline{\lambda}}(t+T, T) \Phi_{\underline{\lambda}}(T, 0) &= \Phi_{\underline{\lambda}}(t+T, T) Q. \\ &= \Phi_{\underline{\lambda}}(t, 0) Q. \end{aligned}$$

$$\Rightarrow \Phi_{\underline{\lambda}}(t+T, 0) = \Phi_{\underline{\lambda}}(t, 0) Q.$$



A very confusing and possibly nontrivial

Lemma :-

Let  $Q$  be a non-singular matrix

$\exists$  a possibly complex matrix  $F$ :

$$e^{FT} = Q.$$

$$P(t) = e^{F(t-t_0)} P(t_0) \Phi_{\Delta}^{-1}(t, t_0)$$

We define  $P(t)$  as above

or

$$P(t) = e^{Ft} \Phi_{\Delta}^{-1}(t, 0)$$

Starting from

$$\dot{x}(t) = A(t)x(t), \quad x(t_0) = x_0$$

$$A(t+T) = A(t) \quad \forall t.$$

Define

$$z(t) = P(t)x(t)$$

$$= \cancel{P(t)} \cancel{P^{-1}(t, t_0)} \cancel{P(t_0)} x(t)$$

$$= e^{F(t-t_0)} P(t_0) \cancel{\Phi_X^{-1}(t, t_0)} x(t)$$

$$\begin{array}{ccc}
 \text{x space } \mathbb{R}^n & \xrightarrow{\Phi_X(t, t_0)} & \mathbb{R}^n \\
 \downarrow P(t_0) & & \downarrow P(t) \\
 \text{z space } \mathbb{R}^n & \xrightarrow{e^{F(t-t_0)}} & \mathbb{R}^n
 \end{array}$$

We obtain  $\dot{z} = \cancel{A} F z, \quad z(t_0) = z_0.$

Claim

$$P^{-1}(t+T) = P^{-1}(t) \quad \forall t.$$

Proof:

$$\begin{aligned}
P^{-1}(t+T) &= \Phi_{\underline{x}}(t+T, 0) e^{-F(t+T)} \\
&= \bar{\Phi}_{\underline{x}}(t+T, T) \Phi_{\underline{x}}(T, 0) e^{-FT} e^{-Ft}. \\
&= \bar{\Phi}_{\underline{x}}(t, 0) e^{FT} e^{-FT} e^{-Ft} \\
&= \bar{\Phi}_{\underline{x}}(t, 0) e^{-Ft} \\
&= P^{-1}(t)
\end{aligned}$$

Remark:

It follows from this claim that

$$z(t) = P(t) x(t)$$

is a Liapunov transformation.

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## Theorem (Floquet - Liapunov)

The dynamical system

$$\dot{x}(t) = A(t)x(t)$$

$A(t)$  is continuous and differentiable in  $[0, \infty)$

$$A(t+T) = A(t)$$

is reducible to a dynamical system with constant co-efficient

$$\dot{z}(t) = Fz(t)$$

via a Liapunov transformation

$$z(t) = P(t)x(t)$$

where

$$P(t) = e^{Ft} \underline{\Phi}_{\underline{x}}^{-1}(t, 0).$$

and where  $F$  is defined as follows

$$e^{FT} \triangleq \underline{\Phi}_{\underline{x}}(T, 0).$$