

# Matrix Exponential.

Introduction: Consider the ~~DE~~ IVP

$$\begin{cases} y' = ay, \\ y(0) = y_0. \end{cases}$$

It has solution  $y(t) = e^{ta} y_0$ . (e.g. solved by separation of variables)

Want to define a matrix  $e^{tA}$ , so that the solution to the IVP

$$\begin{cases} Y' = AY \\ Y(0) = Y_0 \end{cases}$$

(here  $Y = \begin{pmatrix} x(t) \\ y(t) \end{pmatrix}$ ,  $A = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$ )

is given by  $Y(t) = e^{tA} \cdot Y_0$ .

↑ matrix multiplication.

Since the exponential function  $e^x$  admits a Taylor expansion

$$e^x = 1 + x + \frac{x^2}{2!} + \dots + \frac{x^n}{n!} + \dots \quad (n! = n(n-1)(n-2) \dots 2 \cdot 1)$$

we define:

Def: Given a matrix  $B$ , define  $e^B$  as

$$e^B = I + B + \frac{B^2}{2!} + \dots + \frac{B^n}{n!} + \dots$$

identity matrix  
 $I = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$

e.g.  $B^2 = B \cdot B$  (matrix multiplications)

Key property:  $(*) \quad \boxed{\frac{d}{dt} e^{tA} = A \cdot e^{tA}}$

component-wise differentiation = i.e. if  $C(t) = \begin{pmatrix} a(t) & b(t) \\ c(t) & d(t) \end{pmatrix}$ ,

then  $\frac{d}{dt} C(t) = \begin{pmatrix} a'(t) & b'(t) \\ c'(t) & d'(t) \end{pmatrix}$ .

Proof of (\*):

$$\frac{d}{dt} e^{tA} = \frac{d}{dt} \left( I + tA + \frac{t^2 A^2}{2!} + \dots + \frac{t^n A^n}{n!} + \dots \right)$$

$$= \frac{d}{dt} I + \frac{d}{dt} (tA) + \frac{d}{dt} \left( \frac{t^2 A^2}{2!} \right) + \dots + \frac{d}{dt} \left( \frac{t^n A^n}{n!} \right) + \dots$$

(Power rule)

$$= 0 + A + (2t) \frac{A^2}{2!} + \dots + (nt^{n-1}) \frac{A^n}{n!} + \dots$$

$$= A + \frac{tA^2}{1!} + \dots + \frac{t^{n-1} A^n}{(n-1)!} + \dots \quad \left( \text{Note } \frac{n}{n!} = \frac{1}{(n-1)!} \right)$$

$$= A \cdot I + A \cdot (tA) + \dots + A \cdot \left( \frac{t^{n-1} A^{n-1}}{(n-1)!} \right) + \dots$$

$$= A \left( I + tA + \dots + \frac{t^{n-1} A^{n-1}}{(n-1)!} + \dots \right)$$

$$= A \cdot e^{tA} \quad \square$$

From (\*) we can conclude:  $Y(t) = e^{tA} \cdot Y_0$  satisfies

$$Y' = AY =$$

Proof:  $\frac{d}{dt} Y = \frac{d}{dt} (e^{tA} \cdot Y_0) = \left( \frac{d}{dt} e^{tA} \right) \cdot Y_0$  ( $Y_0$  is independent of  $t$ )

$$\stackrel{(*)}{=} A e^{tA} \cdot Y_0$$

$$= AY \quad \square$$

Together with the fact ( $t=0$ )

$$\cancel{Y(0)} = e^{0A} = I + (0A) + \frac{(0A)^2}{2!} + \dots + \frac{(0A)^n}{n!} + \dots \\ = I.$$

$$\Rightarrow Y(0) = e^{0A} \cdot Y_0 = I \cdot Y_0 = Y_0,$$

We have shown that  $Y(t) = e^{tA} \cdot Y_0$  solves the IVP

$$\begin{cases} Y' = AY, \\ Y(0) = Y_0. \end{cases}$$

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Question = how to calculate  $e^{tA}$  explicitly? Note that it's hard to compute  $e^B$  of a matrix directly from the definition = e.g. if  $B = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$ , then

$$B^2 = B \cdot B = \begin{pmatrix} a & b \\ c & d \end{pmatrix} \begin{pmatrix} a & b \\ c & d \end{pmatrix} \\ = \begin{pmatrix} a^2 + bc & ab + bd \\ ac + cd & bc + d^2 \end{pmatrix}$$

is already quite complicated.

To calculate  $e^B$  effectively, we use two observations:

(I) If  $\Lambda = \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix}$  is diagonal, then

$$e^{-\Lambda} = \begin{pmatrix} e^{\lambda_1} & 0 \\ 0 & e^{\lambda_2} \end{pmatrix}.$$

Proof of ①: Note that  $\Lambda^2 = \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix} \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix}$

$$= \begin{pmatrix} \lambda_1^2 & 0 \\ 0 & \lambda_2^2 \end{pmatrix}$$

and in general  $\Lambda^n = \begin{pmatrix} \lambda_1^n & 0 \\ 0 & \lambda_2^n \end{pmatrix}$

$$\Rightarrow e^{-\Lambda} = I + \Lambda + \frac{\Lambda^2}{2!} + \dots + \frac{\Lambda^n}{n!} + \dots$$

$$= \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} + \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix} + \frac{1}{2!} \begin{pmatrix} \lambda_1^2 & 0 \\ 0 & \lambda_2^2 \end{pmatrix} + \dots + \frac{1}{n!} \begin{pmatrix} \lambda_1^n & 0 \\ 0 & \lambda_2^n \end{pmatrix} + \dots$$

$$= \begin{pmatrix} 1 + \lambda_1 + \frac{\lambda_1^2}{2} + \dots + \frac{\lambda_1^n}{n!} + \dots & 0 \\ 0 & 1 + \lambda_2 + \frac{\lambda_2^2}{2!} + \dots + \frac{\lambda_2^n}{n!} + \dots \end{pmatrix}$$

$$= \begin{pmatrix} e^{\lambda_1} & 0 \\ 0 & e^{\lambda_2} \end{pmatrix} \quad \square$$

② If  $A = S\Lambda S^{-1}$ , then  $e^A = S e^{\Lambda} S^{-1}$ .

(Here  $S^{-1}$  is the inverse matrix of  $S$ )

Proof of ②: Note that if  $A = S\Lambda S^{-1}$ , then

$$A^2 = (S\Lambda S^{-1})(S\Lambda S^{-1})$$

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

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

$$= S\Lambda^2 S^{-1}.$$



Similarly  $A^n = S \Lambda^n S^{-1}$

$$\begin{aligned} \Rightarrow e^A &= I + A + \frac{A^2}{2!} + \dots + \frac{A^n}{n!} + \dots \\ &= S I S^{-1} + S \Lambda S^{-1} + S \frac{\Lambda^2}{2!} S^{-1} + \dots + S \frac{\Lambda^n}{n!} S^{-1} + \dots \\ &= S \left( I + \Lambda + \frac{\Lambda^2}{2!} + \dots + \frac{\Lambda^n}{n!} + \dots \right) S^{-1} \\ &= S e^{\Lambda} S^{-1} \quad \square \end{aligned}$$

Thus, using (I), (II), if one can find  $S$  and  $\Lambda = \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix}$ , so that  $A = S \Lambda S^{-1}$ , then

$$e^A \underset{\text{(II)}}{=} S e^{\Lambda} \underset{\text{(I)}}{=} S \begin{pmatrix} e^{\lambda_1} & 0 \\ 0 & e^{\lambda_2} \end{pmatrix} S^{-1}$$

can be computed.

To find  $S$  and  $\Lambda$ , we recall some linear algebra:

Def (Eigenvalue, eigenvector) A number  $\lambda$  is called an eigenvalue of a matrix  $A$ , if there is a non-zero vector  $V$  so that

$$AV = \lambda V.$$

In this case  $V$  is called an eigenvector with respect to  $\lambda$ .

If  $\lambda_1, \lambda_2$  are two eigenvalues of  $A$ , so that the corresponding eigenvectors  $V_1, V_2$  is linearly independent, then

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

with  $S = (V_1 \ V_2)$  and  $\Lambda = \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix}$ .

↑  
Putting 2 column vectors together to form the  $2 \times 2$  matrix  $S$ .

Reason:  $AV_1 = \lambda_1 V_1$ ,  $AV_2 = \lambda_2 V_2$

$$\Rightarrow (AV_1 \ AV_2) = (\lambda_1 V_1 \ \lambda_2 V_2)$$

$$\Rightarrow A(V_1 \ V_2) = (V_1 \ V_2) \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix}$$

$$\Rightarrow AS = S\Lambda$$

( $S$  invertible)

$$\Rightarrow A = S\Lambda S^{-1}$$

To sum up = To calculate  $e^A$ ,

(i) Find its eigenvalues,  $\lambda_1, \lambda_2$

(ii) Find the corresponding eigenvectors,  $V_1, V_2$ .

$$(iii) \Rightarrow e^A = (V_1 \ V_2) \begin{pmatrix} e^{\lambda_1} & 0 \\ 0 & e^{\lambda_2} \end{pmatrix} (V_1 \ V_2)^{-1}$$

Example: Calculate  $e^A$  for  $A = \begin{pmatrix} 5 & 6 \\ -1 & -2 \end{pmatrix}$

To find the eigenvalues, set

$$0 = \det(A - \lambda I) \leftarrow \text{Characteristic polynomial.}$$

$$= \det \begin{pmatrix} 5-\lambda & 6 \\ -1 & -2-\lambda \end{pmatrix}$$

$$= (5-\lambda)(-2-\lambda) + 6$$

$$= \lambda^2 - 3\lambda - 10 + 6$$

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

$$= (\lambda+1)(\lambda-4).$$

$$\Rightarrow \lambda_1 = -1, \lambda_2 = 4.$$

To find the eigenvectors,

$$\underline{\lambda_1 = -1}: A - \lambda_1 I = \begin{pmatrix} 5 & 6 \\ -1 & -2 \end{pmatrix} + \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 6 & 6 \\ -1 & -1 \end{pmatrix}$$

$$\Rightarrow v_1 = \begin{pmatrix} 1 \\ -1 \end{pmatrix} \text{ is one eigenvector.}$$

$$\underline{\lambda_2 = 4}: A - \lambda_2 I = \begin{pmatrix} 1 & 6 \\ -1 & -6 \end{pmatrix} \Rightarrow v_2 = \begin{pmatrix} 6 \\ -1 \end{pmatrix}.$$

$$\Rightarrow S = (v_1 \ v_2) = \begin{pmatrix} 1 & 6 \\ -1 & -1 \end{pmatrix} \Rightarrow S^{-1} = \frac{1}{-1+6} \begin{pmatrix} -1 & -6 \\ 1 & 1 \end{pmatrix} \\ = \begin{pmatrix} -\frac{1}{5} & -\frac{6}{5} \\ \frac{1}{5} & \frac{1}{5} \end{pmatrix}.$$

$$\Rightarrow e^A = \begin{pmatrix} 1 & 6 \\ -1 & -1 \end{pmatrix} \begin{pmatrix} e^{-1} & 0 \\ 0 & e^4 \end{pmatrix} \begin{pmatrix} -\frac{1}{5} & -\frac{6}{5} \\ \frac{1}{5} & \frac{1}{5} \end{pmatrix} \quad \square$$