

## Discussion 6D

## MT2 Review

\* Linearization

\* Discretization

\* Stability

\* Feedback Control (up to Tuesday's lecture:  
no CCF, or tracking control)

### (I) Linearization

Everything comes from Taylor expansions

$$f(x) = \sum_{n=0}^{\infty} \frac{f^{(n)}(x^*)}{n!} (x-x^*)^n$$

To linearize, take terms up to order 1

$$\begin{aligned} f(x) &\approx f(x^*) + f'(x^*) (x-x^*) \\ \delta f &= f(x) - f(x^*) \\ \delta x &= x - x^* \end{aligned}$$

Scalar  
case

In general, vector case

$$f(\vec{x}) \approx f(\vec{x}^*) + \left. \nabla f \right|_{\vec{x}=\vec{x}^*} (\vec{x} - \vec{x}^*)$$

where  $\nabla f$  denotes the Jacobian

| - - - - - |

$$\nabla f = \begin{pmatrix} \frac{\partial f_1}{\partial x_1} & \dots & \frac{\partial f_1}{\partial x_n} \\ \vdots & & \vdots \\ \frac{\partial f_m}{\partial x_1} & \dots & \frac{\partial f_m}{\partial x_n} \end{pmatrix}$$

$$\vec{x} - \vec{x}^* = \begin{bmatrix} x_1 - x_1^* \\ \vdots \\ x_n - x_n^* \end{bmatrix}$$

For some state space model, we have:

$$\frac{d\vec{x}}{dt} = f(\vec{x}, \vec{u}) \quad \text{CT}$$

$$\text{or } \vec{x}[t+1] = f(\vec{x}[t], \vec{u}[t]) \quad \text{DT}$$

$$f(\vec{x}, \vec{u}) \approx f(\vec{x}^*, \vec{u}^*) + \nabla_x f \Big|_{\vec{x}^*, \vec{u}^*} (\vec{x} - \vec{x}^*) + \nabla_u f \Big|_{\vec{x}^*, \vec{u}^*} (\vec{u} - \vec{u}^*)$$

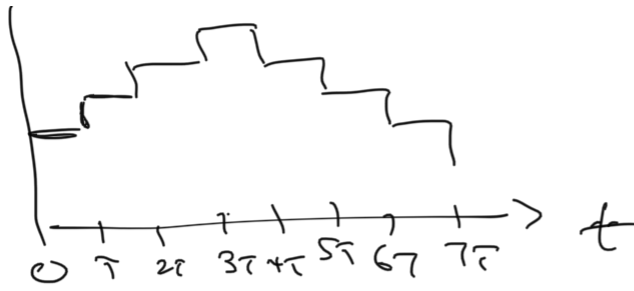
$\swarrow$   $\vec{x}^*, \vec{u}^*$   $\swarrow$   $\vec{x}^*, \vec{u}^*$   
 $A$   $B$

$$A\vec{x} + B\vec{u}$$

## (II) Discretization

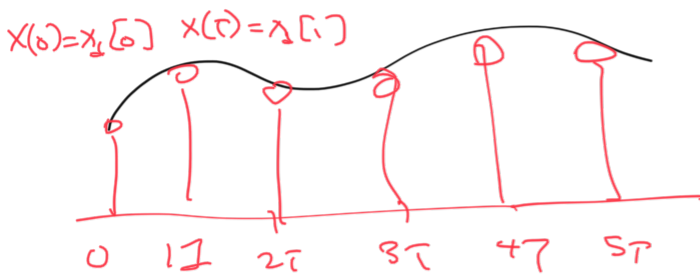
Idea: use digital control, which in CT we represent as piecewise constant inputs

$$\rightarrow u(t)$$



Notation:  $x_d[k] = x[kT] \leftarrow CT$   
 $\uparrow$   
 $\Delta T$

$$u_d[k] = u[kT]$$



Method: solve differential equations

$$\frac{dx}{dt} = \lambda x(t) + bu(t)$$

$$\text{or } \frac{dx}{dt} = Ax(t) + Bu(t)$$

with a constant input

in the interval  $t \in [kT, (k+1)T)$



In this time interval,  $u(t) = u_d[k] = \text{const.}$

Using techniques from the 1st half of class:

Scalar case

$$x(t) = e^{\lambda(t-kT)} x(kT) + b u_d[k] \int_0^T e^{-\lambda s} ds$$

∴

$$t = (k+1)T$$

Plug in  $u_d$

$$x((k+1)T) = e^{AT} x(kT) + b u_d[k] \int_0^T e^{A_s} ds$$

$$\boxed{x_d[k+1] = e^{-dT} x_d[k] + \left( b \int_0^T e^{A_s} ds \right) u_d[k]}$$

$\uparrow$   $\quad$   $\uparrow$   
 $A_d$   $\quad$   $b_d$

• Vector case

↳ Try to turn into a system of independent scalar case

a)  $A$  is diagonalizable

$$\frac{d\vec{x}}{dt} = A\vec{x} + B\vec{u}$$

$$\vec{z} = V^{-1}\vec{x}$$

$$\frac{d\vec{z}}{dt} = \Lambda\vec{z} + V^{-1}B\vec{u}$$

After even more math:

$$\vec{z}_d[k+1] = e^{\Lambda T} \vec{z}_d[k] + \Lambda_T V^{-1} B \vec{u}_d[k]$$

Go back to original basis, convert  $\forall \vec{x} = V\vec{z}$

$$\boxed{\vec{x}_d[k] = V e^{\Lambda T} V^{-1} \vec{x}_d[k] + V \Lambda_T V^{-1} B \vec{u}_d[k]}$$



where

$$e^{\Lambda T} = \begin{bmatrix} e^{\lambda_1 T} & & 0 \\ & \dots & \\ 0 & & e^{\lambda_n T} \end{bmatrix}$$

$$\Lambda_T = \begin{bmatrix} \int_0^T e^{\lambda_1 s} ds & & 0 \\ & \dots & \\ 0 & & \int_0^T e^{\lambda_n s} ds \end{bmatrix}$$

b)  $A$  is not diagonalizable

- matrix exponential (out-of-scope)

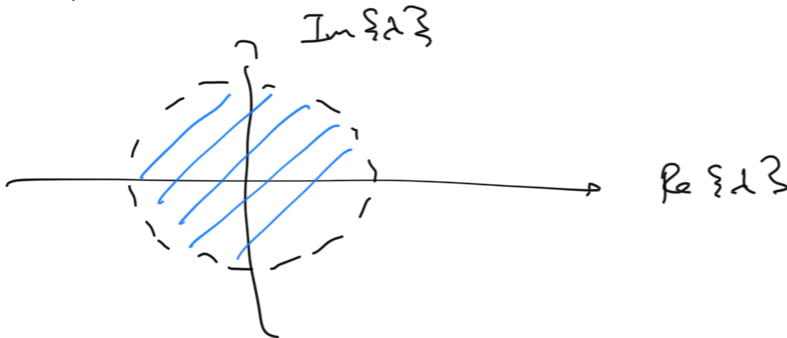
- solve diff eq w/ const inputs

using other means if system is simple enough

(e.g. car model: used FTC and directly integrated)

### III. Stability

a) Discrete Time



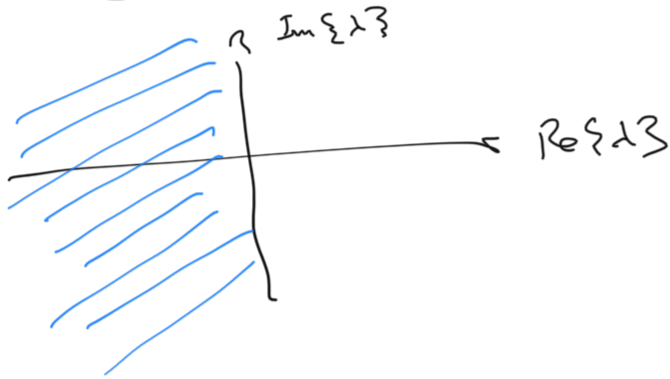
Stable:  $|x_i| < 1 \quad \forall i$

unstable:  $\exists x_i$  s.t.  $|x_i| > 1$

marginally stable?  $|x_i| \leq 1 \quad \forall i$  and  $\exists \lambda_i$  s.t.  $|\lambda_i| = 1$

marginaly

b) Continuous Time



stable:  $\text{Re}\{\lambda_i\} < 0 \quad \forall i$

unstable:  $\exists \lambda_i \text{ s.t. } \text{Re}\{\lambda_i\} > 0$

marginaly stable:  $\text{Re}\{\lambda_i\} \leq 0 \quad \forall i$  and  $\exists |\lambda_j| \text{ s.t. } \text{Re}\{\lambda_j\} = 0$

Added note on marginal stability:

$$x(t) = x_0 e^{\lambda t} + b \int_0^t e^{\lambda(t-\tau)} u(\tau) d\tau$$

Input  $u(t) = e^{j\omega_0 t}$  where  $\lambda = j\omega_0$

(so match the frequency of the eigenvalue)



$\Rightarrow$  pure resonance

bounded input, unbounded output

$$x(t) = x_0 e^{j\omega_0 t} + b \int_0^t e^{j\omega_0 t} \cancel{e^{-j\omega_0 \tau}} e^{j\omega_0 \tau} d\tau$$

$$= \dots + b e^{j\omega_0 t} \int_0^t d\tau$$

$$\lim_{t \rightarrow \infty} x(t) \rightarrow \infty$$

### III. Feedback Control

open-loop:  $u[k]$  independent of outputs

closed-loop:  $u[k]$  dependent on outputs,

i.e.  $u[k] = f(x[k])$

Typically, linear control:

$$\begin{array}{c} \vec{u}[k] = -K \vec{x}[k] \\ \uparrow \qquad \qquad \uparrow \qquad \uparrow \\ m \times 1 \qquad m \times n \qquad n \times 1 \end{array}$$

Then, 
$$\begin{aligned} \hat{x}[k+1] &= A\hat{x}[k] + B u[k] \\ &= A\hat{x}[k] - BK\hat{x}[k] \\ &= \underbrace{(A - BK)}_{A_{cl}} \hat{x}[k] \end{aligned}$$

If system is controllable, guaranteed we can place the eigenvalues anywhere

(CCF)

↑  
not on Tues  
lecture,  
ergo it's  
not in

Choose eigenvalues and get same  
desired char. poly:  $\chi_{goal}$



desired  $\lambda$  values

⇒ match  $\lambda_{goal}$  to  $\lambda_{Acl}$

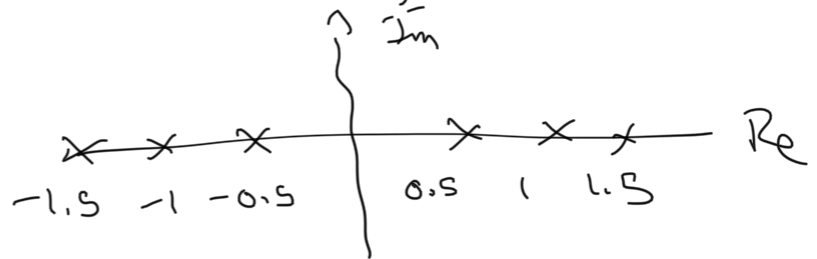
function of  $K$

Match coefficients to choose  $K$  values

① Discrete, Q1 — DT System Response

$$x[k+1] = \lambda x[k]$$

$$x[0] = 1$$



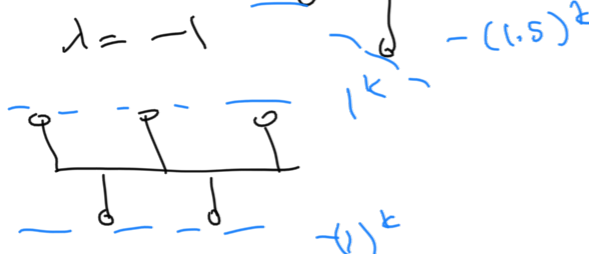
(i)  $\lambda = -1.5$

$$x[k] = \lambda^k x[0] = (-1.5)^k x[0] = (-1.5)^k$$



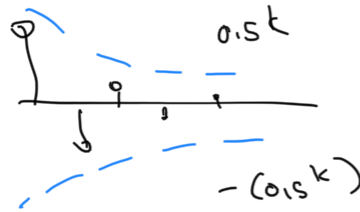
unstable,  
b/c  $| -1.5 | > 1$

(ii)  $\lambda = -1$



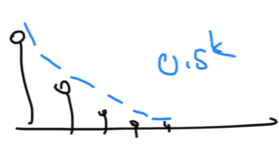
marginally stable

(iii)  $\lambda = -0.5$



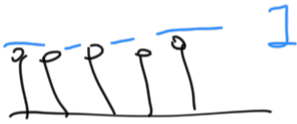
stable,  
b/c  $| -0.5 | < 1$

(iv)  $\lambda = 0.5$



stable,  
b/c  $| 0.5 | < 1$

iv)  $\lambda = 1$



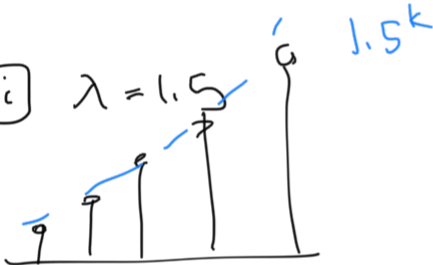
marginally stable

$x[0] = 1$

$\lambda = j$   
 $x[k] = j^k x[0] = j^k$



vi)  $\lambda = 1.5$



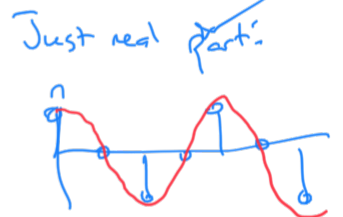
unstable,  
 b/c  $|\lambda| > 1$

② Discrete, Q3

Eigenvalue Placement in CCF

$$\bar{x}[k+1] = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & -9 & -6 \end{bmatrix} \bar{x}[k] + \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \bar{u}[k]$$

A B



Just real part:  
 Recall that  
 $|\lambda|^k e^{j\omega k}$   
 $j = 1 e^{j\frac{\pi}{2}}$   
 $e^{j\frac{\pi}{2}k}$   
 Real part:  
 $x[k] = \cos(\frac{\pi}{2}k)$

a)

$$\det(A - \lambda I) = \det \begin{vmatrix} \lambda & 1 & 0 \\ 0 & -\lambda & 1 \\ 0 & -9 & -6-\lambda \end{vmatrix}$$

$$= -\lambda \begin{vmatrix} -\lambda & 1 \\ -9 & -6-\lambda \end{vmatrix} - 1 \begin{vmatrix} 0 & 1 \\ 0 & -6-\lambda \end{vmatrix} + 0 \begin{vmatrix} 0 & -\lambda \\ 0 & -9 \end{vmatrix}$$

$$= -\lambda (\lambda(\lambda+6) + 9) = -\lambda (\lambda^2 + 6\lambda + 9) = 0$$

$$\Rightarrow \lambda(\lambda+3)^2 = 0$$

$$\lambda = 0, \underset{\uparrow}{-3}, -3$$



$$-6 + k_2 = 0 \implies k_2 = 6$$

$$9 - k_1 = -\frac{1}{4} \implies k_1 = 9 + \frac{1}{4} = \frac{37}{4}$$

$$k_0 = 0$$

$$K = [k_0 \ k_1 \ k_2] = \left[ 0 \ \frac{37}{4} \ 6 \right]$$

If we used  $u[k] = -Kx[k]$

Then  $K = \left[ 0 \ -\frac{37}{4} \ -6 \right]$  (Just flip by a minus sign)

"Placing eigenvalues"

$A$  has some eigenvalues — potentially unstable

$A_{cl}$  has eigenvalues that depend on  $K$  — has eigenvalues that depend on  $K$

We choose  $K$  s.t. the eigenvalues of  $A_{cl}$  (design)

are where we want them to be (e.g. make all  $|z_i| < 1$  to make stable)

c) Consider the case that  $k_i$  are limited to

$$[-B, B]$$

Can we make the system stable?



← very edge of where



all the  $\lambda$  could be  
and still be stable

Suppose  $\lambda = -1$  : then  $\chi = (\lambda + 1)^3$   
 $= \lambda^3 + 3\lambda^2 + 3\lambda + 1$

$\lambda = 1$  : then  $\chi = (\lambda - 1)^3$   
 $= \lambda^3 - 3\lambda^2 + 3\lambda - 1$

$\lambda = -j$  : then  $\chi = (\lambda + j)^3 = \lambda^3 + 3j\lambda^2 - 3\lambda - j$

Then the magnitude of the coefficients has to be:

$|c_0| = 1$     $|c_1| = 3$  ,    $|c_2| = 3$  ,    $|c_3| = 1$

$\uparrow$                        $\uparrow$                        $\uparrow$                        $\uparrow$   
 $\lambda^0$                        $\lambda^1$                        $\lambda^2$                        $\lambda^3$

Try  $\lambda = -0.99$  ,                       $\chi = c_0 + c_1\lambda + c_2\lambda^2 + c_3\lambda^3$

then  $(\lambda + 0.99)^3 = \lambda^3 + 2.97\lambda^2 + 2.9403\lambda + 0.970299$  ...

$\uparrow$                        $\uparrow$                        $\uparrow$                        $\uparrow$   
 $|c_3| = 1$                $|c_2| < 3$                $|c_1| < 3$                $|c_0| < 1$

Conditions on coefficients for  $\lambda$  to all be  
inside the unit circle

⇒ Compare to  $\chi_{ACL}$

$\lambda^3 - (-6+k_2)\lambda^2 - (-9+k_1)\lambda - k_0$

$\uparrow$                        $\uparrow$                        $\uparrow$

fine  $k_2 = 3$ ,  
fine

$k_1 = 6$ ,  
out of range!

Can never make the system stable!

③ Fall 2011, MT 2, Q 2

Linearization + Stability + Equilibrium Points

$$x[t+1] = f(x[t])$$

$$f(x) = 2x - 2x^2$$

a) Equilibrium points:

b) Linearize about each eq. point, Stable?

a) DT system:

equilibrium  $\rightarrow x[t+1] = x[t]$  ← constant solutions

$$f(x) = x$$

$$\Rightarrow 2x - 2x^2 = x$$

$$x - 2x^2 = 0$$

$$x(1 - 2x) = 0$$

$$x^* = 0, 0.5$$

1) ...

$$b) \quad \lambda = 0,$$

Scalar system!

$$f(x) \approx f(x^*) + f'(x^*) (x - x^*)$$

$$f(x) \approx f(0) + f'(0) (x)$$

$$f(0) = 2(0) - 2(0^2) = 0$$

$$f'(0) = (2 - 4x) \Big|_{x=0} = 2$$

Thus,  $f(x) \approx 2x$

Stable?

$$x[t+1] = f(x[t]) \approx 2x[t] = \lambda x[t] + b_u[t]$$

$$\lambda = 2 > 1 \implies \text{unstable}$$

$$x^* = 0.5$$

$$f(x) \approx f(0.5) + f'(0.5) (x - 0.5)$$

$$f(0.5) = 2(0.5) - 2(0.5^2) = 0.5$$

$$f'(0.5) = 2 - 4(0.5) = 0$$

$$f(x) = 0.5$$

$$x[t+1] = f(x[t]) = 0.5 = \lambda x[t] + b_u[t]$$

$$\implies \lambda = 0,$$

(stable)

(4) Discrete, Q1, part b

$$\frac{dx}{dt} = \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix} x(t) + \begin{bmatrix} 0 \\ 2 \end{bmatrix} u(t)$$

$$\Rightarrow \vec{x}[k+1] = \underline{A}_d \vec{x}[k] + \underline{B}_d u[k]$$

$$\left. \begin{aligned} \underline{A}_d &= V \Lambda^T V^{-1} \\ \underline{B}_d &= V \Lambda^T V^{-1} B \end{aligned} \right\}$$

First confirm you have a diagonalizable system!

$$\det(A - \lambda I) = \begin{vmatrix} -\lambda & 1 \\ -2 & -3-\lambda \end{vmatrix} = \lambda(\lambda+3) + 2 \\ = \lambda^2 + 3\lambda + 2 \\ = (\lambda+1)(\lambda+2)$$

$$\Rightarrow \lambda_1 = -2 \\ \lambda_2 = -1$$

$$\underline{\Lambda} = \begin{bmatrix} -2 & 0 \\ 0 & -1 \end{bmatrix}$$

Wouldn't work for car model, for example

$$\begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \leftarrow \text{not diagonalizable}$$

$$A + 2I = \begin{bmatrix} 2 & 1 \\ -2 & -1 \end{bmatrix} \rightarrow \vec{v}_1 = \begin{bmatrix} 1 \\ -2 \end{bmatrix}$$

$$A + 1I = \begin{bmatrix} 1 & 1 \\ -2 & -2 \end{bmatrix} \rightarrow \vec{v}_2 = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

$$\dots \begin{bmatrix} 1 & 1 \end{bmatrix}$$

$$\dots \begin{bmatrix} 1 & -1 \end{bmatrix} \quad \begin{bmatrix} 1 & -2 \end{bmatrix}$$



$$V = \begin{bmatrix} 1 & 1 \\ -2 & -1 \end{bmatrix} \quad \begin{bmatrix} a & b \\ c & d \end{bmatrix} = \frac{1}{ad-bc} \begin{bmatrix} -d & b \\ c & -a \end{bmatrix}$$

$$V^{-1} = \frac{1}{-1+2} \begin{bmatrix} -1 & -1 \\ 2 & 1 \end{bmatrix} = \begin{bmatrix} -1 & -1 \\ 2 & 1 \end{bmatrix}$$

$$A_d = V e^{\Lambda T} V^{-1} = \begin{bmatrix} 1 & 1 \\ -2 & -1 \end{bmatrix} \begin{bmatrix} e^{-2T} & 0 \\ 0 & e^{-T} \end{bmatrix} \begin{bmatrix} -1 & -1 \\ 2 & 1 \end{bmatrix}$$

$$= \begin{bmatrix} 1 & 1 \\ -2 & -1 \end{bmatrix} \begin{bmatrix} -e^{-2T} & -e^{-2T} \\ 2e^{-T} & e^{-T} \end{bmatrix}$$

$$A_d = \begin{bmatrix} -e^{-2T} + 2e^{-T} & -e^{-2T} + e^{-T} \\ -2e^{-2T} - 2e^{-T} & 2e^{-2T} - e^{-T} \end{bmatrix}$$

What is  $T$ ?

Recall that the inputs are piecewise constant:



$$B_d = V \Lambda_T V^{-1} B$$

$$\Lambda_T = \begin{bmatrix} \int_0^T e^{\lambda_1 s} ds & 0 \\ 0 & \int_0^T e^{\lambda_2 s} ds \end{bmatrix}$$

$$= \begin{bmatrix} \int_0^T e^{-2s} ds & 0 \\ 0 & \int_0^T e^{-s} ds \end{bmatrix}$$

$$= \begin{bmatrix} \frac{e^{-2s}}{-2} \big|_{0^+} & 0 \\ 0 & \frac{e^{-s}}{-1} \big|_{0^+} \end{bmatrix}$$

$$= \begin{bmatrix} \frac{e^{-2T}-1}{-2} & 0 \\ 0 & \frac{e^{-T}-1}{-1} \end{bmatrix}$$

$$\begin{bmatrix} 1 & 1 \\ -2 & -1 \end{bmatrix} \begin{bmatrix} \frac{e^{-2T}-1}{-2} & 0 \\ 0 & \frac{e^{-T}-1}{-1} \end{bmatrix} \begin{bmatrix} -1 & -1 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} 0 \\ 2 \end{bmatrix}$$

$$= \begin{bmatrix} 1 & 1 \\ -2 & -1 \end{bmatrix} \begin{bmatrix} \frac{e^{-2T}-1}{-2} & 0 \\ 0 & \frac{e^{-T}-1}{-1} \end{bmatrix} \begin{bmatrix} -2 \\ 2 \end{bmatrix}$$

$$= \begin{bmatrix} 1 & 1 \\ -2 & -1 \end{bmatrix} \begin{bmatrix} e^{-2T}-1 \\ -2(e^{-T}-1) \end{bmatrix}$$

$$\Rightarrow B_d = \begin{bmatrix} e^{-2T}-1 - 2(e^{-T}-1) \\ -2(e^{-2T}-1) + 2(e^{-T}-1) \end{bmatrix} = \begin{bmatrix} 1 - 2e^{-T} + e^{-2T} \\ 2e^{-T} - 2e^{-2T} \end{bmatrix}$$