## Stability of Discrete-Time LTI Systems: Solving Exercises in MATLAB

In this live script we illustrate, by examples, how to use MATLAB when solving stability and state movement exercises for discrete-time LTI systems.

## **Exercise 1: Stability & Equilibrium State Analysis**

Given the LTI system

$$\begin{cases} x(n+1) = \begin{bmatrix} 0 & +1 \\ -1 & 0 \end{bmatrix} x(n) + \begin{bmatrix} -1 \\ +1 \end{bmatrix} u(n) \\ y(n) = \begin{bmatrix} +2 & -2 \end{bmatrix} x(n) \end{cases}$$

Q1:

- Determine the characteristic polynomial of the system and the eigenvalues of the matrix *A*.
- · Using the eigenvalues, what can be state about the stability of the system?

```
clear
close all
clc
% let's clear the workspace
```

Let's assign the Amatrix (we will use the matrices A, B, C, D later)

```
A = [0, +1; ... \\ -1, 0];
```

and compute the characteristic polynomial  $p_A(\lambda) = \det(\lambda I - A)$ 

```
pA = tv3
```

$$pA = 1 \times 3$$
  
1 0 1

Please note: A is a numeric matrix, so the command charpoly provides an array with the polynomial coefficients. Indeed, the polynomial is

$$p_A(\lambda) = +1 \ \lambda^2 + 0 \ \lambda + 1 = \lambda^2 + 1$$

The eigenvalues are the roots of the polynomial  $p_A(\lambda)$ . It is easy to evaluate numerically the roots of a polynomial, given the array of the polynomial coefficients

```
eigA = roots(pA)
```

```
eigA = 2×1 complex
0.0000 + 1.0000i
0.0000 - 1.0000i
```

The eigenvalues are two complex conjugate values:  $\lambda_{A,1} = 0 + 1i$ ,  $\lambda_{A,2} = 0 - 1i$ .

What about the eigenvalues magnitude?

```
eigMAG = abs(eigA)
eigMAG = 2×1
1
```

Of course, both the eigenvalues have <u>unitary magnitude</u>, and they are <u>distinct eigenvalues</u>. This result allows us to state: **the system is marginally stable**.

What about the state free movement? What are the response mode matrices?

Rember, the eigenvalues are all distinct, so we could use the formula

$$A^{k} = \sum_{i=1}^{n} A_{i} \lambda_{i}^{k} \qquad A_{i} = \lim_{z \to \lambda_{i}} \left[ (z - \lambda_{i})(zI - A)^{-1} \right] i = 1, 2, \dots n$$

or exploiting the left and right eigenvector:

$$A_i = v_i \cdot \widetilde{v}_i^{\mathsf{T}}$$

Unfortunately, due to the complex eigenvalues, we can not evaluate the expressions

$$A_i = \lim_{z \to \lambda_i} \left[ (z - \lambda_i)(zI - A)^{-1} \right] \ i = 1 \ , \ 2 \ , \ \dots n$$
 using the Symbolic Math Toolbox. In fact, the command limit can

handle only limits where the limit point is a reeal finite value, or  $+\infty$  either  $-\infty$ . Experiment what happens, if you try to use the command limit - please uncommment the script row 20 and execute this code section

```
syms z n
zIA = z*eye(size(A))-A;
zIAinv = inv(zIA);
lambdaA = eig(sym(A));
% A1 = limit((z-lambdaA(1))*zIAinv,z,lambdaA(1))
```

Using left and right eigenvectors is still possible evaluate the response mode matrices.

```
[R_eigVects, lambdaSET] = eig(sym(A))
```

 $\begin{array}{l} \text{R\_eigVects} = \\ \begin{pmatrix} i & -i \\ 1 & 1 \end{pmatrix} \\ \text{lambdaSET} = \\ \begin{pmatrix} -i & 0 \\ 0 & i \end{pmatrix} \end{array}$ 

% extracting the main diagonal from the square matrix lambdaSET, creating
% a column vector and assigning the col. vector to lambdaSET
lambdaSET = diag(lambdaSET);

And for evaluating the left eigenvectors, remeber

$$Q := [v_1 \mid v_2 \mid \cdots \mid v_n] \implies P = Q^{-1} = \begin{bmatrix} \widetilde{v}_1^\top \\ \vdots \\ \widetilde{v}_n^\top \end{bmatrix}; \widetilde{v}_i^\top v_j = \begin{cases} 1 & i = j \\ 0 & i \neq j \end{cases}$$

Thus

% the left eigenvectors as rows of the inverse matrix, build with the right % eigenvectors as columns  $L\_eigVects = inv(R\_eigVects)$ 

L\_eigVects =

$$\begin{pmatrix} -\frac{1}{2} i & \frac{1}{2} \\ \frac{1}{2} i & \frac{1}{2} \end{pmatrix}$$

Finally, we can compute the response mode matrices, by simply evaluating

A1\_mat = R\_eigVects(:,1)\*L\_eigVects(1,:) %#ok<\*NASGU>

 $A1_mat =$ 

$$\begin{pmatrix}
\frac{1}{2} & \frac{1}{2}i \\
-\frac{1}{2}i & \frac{1}{2}
\end{pmatrix}$$

A2\_mat = R\_eigVects(:,2)\*L\_eigVects(2,:)

 $A2_mat =$ 

$$\begin{pmatrix} \frac{1}{2} & -\frac{1}{2} i \\ \frac{1}{2} i & \frac{1}{2} \end{pmatrix}$$

The matrix power  $A^n$  is

Apow =

$$\begin{pmatrix} \frac{(-i)^n}{2} + \frac{i^n}{2} & \frac{(-i)^n i}{2} - \frac{i^n i}{2} \\ -\frac{(-i)^n i}{2} + \frac{i^n i}{2} & \frac{(-i)^n}{2} + \frac{i^n}{2} \end{pmatrix}$$

Please, verify that you obtain the same result for the matrix  $A^n$  also applying the inverse Z-transform to  $z \cdot (zI - A)^{-1}$ .

## simplify(expand(Apow))

ans =

$$\begin{pmatrix} \frac{(-1)^n + 1}{2 (-1)^{n/2}} & -\frac{(-1)^n i - i}{2 (-1)^{n/2}} \\ \frac{(-1)^n i - i}{2 (-1)^{n/2}} & \frac{(-1)^n + 1}{2 (-1)^{n/2}} \end{pmatrix}$$

$$V^{-1} * A * V = J \Rightarrow A = V * J * V^{-1} \Rightarrow A^{k} = V * J^{k} * V^{-1}$$

```
[V, J] = jordan(A);
%% inv(V) * sysD.A * V = J

k = sym("k");
A_sym = sym(A);
modes = V * A_sym^k * inv(V)
```

modes =

$$\begin{pmatrix} \sigma_2 & \sigma_1 \\ \sigma_1 & \sigma_2 \end{pmatrix}$$

where

$$\sigma_1 = -\frac{e^{-\frac{\pi k i}{2}}}{2} + \frac{e^{\frac{\pi k i}{2}}}{2}$$

$$\sigma_2 = \frac{e^{-\frac{\pi k i}{2}}}{2} + \frac{e^{\frac{\pi k i}{2}}}{2}$$

```
for i=0:10
    a = subs(Apow, n, i);
    b = subs(modes, k, i);
    isequal(a,b)
end
```

```
ans = logical

ans = logical

0

ans = logical

1

ans = logical

0

ans = logical

1

ans = logical

1

ans = logical

0

ans = logical

1

ans = logical

1

ans = logical

0

ans = logical

1

ans = logical

1

ans = logical

1

ans = logical
```

## return

Consider the same LTI system

$$\begin{cases} x(n+1) &= \begin{bmatrix} 0 & +1 \\ -1 & 0 \end{bmatrix} x(n) + \begin{bmatrix} -1 \\ +1 \end{bmatrix} u(n) \\ y(n) &= \begin{bmatrix} +2 & -2 \end{bmatrix} x(n) \end{cases}$$

**Q2**:

- verify that  $\bar{x} = [0 \ 0]^{\top}$  is an equilibrium state for the system, when the input  $u(n) = \bar{u} = 0 \quad \forall n \in \mathbb{Z}$  is applied;
- evaluate the state free movement, obtained starting from an initial state  $\hat{x}$  such that  $||\hat{x} \bar{x}|| \le \frac{1}{10}$ .

Select 10 different initial states  $\hat{x}$ , repeat the computation of the state free movement starting from each of them, store each resulting state sequence, and plot all the 10 movements in a 3D figure (with the time instants n as the first coordinate, the state component  $x_1$  as the second coordinate, and the state component  $x_2$  as the third coordinate). Use different markers, different colours, different line styles and an appropriate description in the graph legend to enable the identification of each individual state movement.

• Assume  $T_s = 1$  as sampling period, and evaluate the state movement from the initial time instant  $t_0 = 0$  to the time instant  $t_{\text{stop}} = 50$ 

Let's verify that  $\bar{x}$  is an equilibrium state, when the input  $\bar{u}$  is applied.

Remember, the pair  $\bar{x}$ ,  $\bar{u}$  corresponds to an equilibrium state and to an equilibrium input if and only if the pair satisfy the algebraic equation

$$\overline{x} = A\overline{x} + B\overline{u} \iff (I - A)\overline{x} = B\overline{u}$$

```
bar_x = [0;0];
bar_u = 0;
res = A*bar_x+B*bar_u
```

So it has been verified!

Generating 10 states  $\hat{x}_{(k)}$   $k = 1, 2, \dots 10$  such that  $||\hat{x} - \overline{x}|| \le \frac{1}{10}$ , where  $\overline{x} = [0 \ 0]^{\mathsf{T}}$ 

```
R = 10; % 10 state vectors x_hat = zeros(2, R); % preallocate the array Nhat_found = 0; % actual number of state vectors in the array x_hat MAX_Nhat = 10; % the requested number of vectors maxDIST = 1e-1; % the maximum distance between bar_x and each vector in x_hat
```

Let us generate random vectors, test the distance from the null state, and each time one of them satisfies the *maximum distance from the origin* constraint, we store it. Let us repeat these operations until we have the required 10 vectors.

Now we are able to evaluate each free movement

Having the state movements, it is possible to plot the results