# Stability Control Theory, Lecture 2

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# CRITICAL POINT (NODE)

Consider the following ODE:

$$\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x}, t) \tag{1}$$

Let  $\mathbf{x}_0$  be such a state that:

$$\mathbf{f}(\mathbf{x}_0, t) = 0 \tag{2}$$

Then such state  $\mathbf{x}_0$  is called a *node* or a *critical point*.

#### STABILITY

Node  $\mathbf{x}_0$  is called *stable* iff for any constant  $\delta$  there exists constant  $\varepsilon$  such that:

$$||\mathbf{x}(0) - \mathbf{x}_0|| < \delta \longrightarrow ||\mathbf{x}(t) - \mathbf{x}_0|| < \varepsilon$$
 (3)

Think of it as "for any initial point that lies at most  $\delta$  away from  $\mathbf{x}_0$ , the rest of the trajectory  $\mathbf{x}(t)$  will be at most  $\varepsilon$  away from  $\mathbf{x}_0$ ".

Or, more picturesque, think of it as "the solutions with different initial conditions do not diverge from the node"

#### Asymptotic stability

Node  $\mathbf{x}_0$  is called *asymptotically stable* iff for any constant  $\delta$  it is true that:

$$||\mathbf{x}(0) - \mathbf{x}_0|| < \delta \longrightarrow \lim_{t \to \infty} \mathbf{x}(t) = \mathbf{x}_0$$
 (4)

Think of it as "for any initial point that lies at most  $\delta$  away from  $\mathbf{x}_0$ , the trajectory  $\mathbf{x}(t)$  will asymptotically approach the point  $\mathbf{x}_0$ ".

Or, more picturesque, think of it as "the solutions with different initial conditions converge to the node"

## STABILITY VS ASYMPTOTIC STABILITY

#### Example

Consider dynamical system  $\dot{x} = 0$ , and solution x = 7. This solution is stable, but not asymptotically stable (other solutions do not diverge from x = 7, but do not converge to it either).

#### Example

Consider dynamical system  $\dot{x} = -x$ , and solution x = 0. This solution is stable and asymptotically stable (other solutions converge to x = 0).

#### Example

Consider dynamical system  $\dot{x} = x$ , and solution x = 0. This solution is unstable (other solutions diverge from x = 0).

#### LTI AND AUTONOMOUS LTI

Consider the following linear ODE:

$$\dot{\mathbf{x}} = \mathbf{A}\mathbf{x} + \mathbf{B}\mathbf{u} \tag{5}$$

This is called a *linear time-invariant system*, or *LTI*.

Consider the following linear ODE:

$$\dot{\mathbf{x}} = \mathbf{A}\mathbf{x} \tag{6}$$

This is also an LTI, but it is also called an *autonomous system*, since its evolution depends only on the state of the system.

Example: real eigenvalues

Consider autonomous LTI:

$$\dot{\mathbf{x}} = \mathbf{A}\mathbf{x} \tag{7}$$

where **A** can be decomposed via eigen-decomposition as  $\mathbf{A} = \mathbf{V}\mathbf{D}\mathbf{V}^{-1}$ , where **D** is a diagonal matrix.

$$\dot{\mathbf{x}} = \mathbf{V}\mathbf{D}\mathbf{V}^{-1}\mathbf{x} \tag{8}$$

Multiply it by  $\mathbf{V}^{-1} \longrightarrow \mathbf{V}^{-1}\dot{\mathbf{x}} = \mathbf{V}^{-1}\mathbf{V}\mathbf{D}\mathbf{V}^{-1}\mathbf{x}$ . Define  $\mathbf{z} = \mathbf{V}^{-1}\mathbf{x} \longrightarrow \dot{\mathbf{z}} = \mathbf{D}\mathbf{z}$ .

Since elements of  $\mathbf{D}$  are real, we can clearly see, that iff they are all negative will the system be asymptotically stable. If they are non-positive, the system is stable. And those elements are eigenvalues of  $\mathbf{A}$ .

Example: complex eigenvalues, part 1

Assume that  $\mathbf{A}$  can be decomposed via eigen-decomposition as  $\mathbf{A} = \mathbf{U}\mathbf{C}\mathbf{U}^{-1}$ , where  $\mathbf{C}$  is a complex-valued diagonal matrix and  $\mathbf{U}$  is a complex-valued inevitable matrix.

We can perform the same steps (multiply by  $\mathbf{U}^{-1}$ , then define  $\mathbf{z} = \mathbf{U}^{-1}\mathbf{x}$ ) to arrive at:

$$\dot{\mathbf{z}} = \mathbf{C}\mathbf{z} \tag{9}$$

which falls into a set of independent equations, with complex coefficients  $c_i$ :

$$\dot{z}_i = c_i z_i \tag{10}$$

The solution is:

$$z_i = k_0 e^{c_i t} \tag{11}$$

Example: complex eigenvalues, part 2

The solution  $z_i = k_0 e^{c_i t}$ , where  $c_i = \alpha_i + i\beta_i$ , can be decomposed using Euler's identity:

$$z_i = k_0 e^{c_i t} = k_0 e^{(\alpha_i + i\beta_i)t} = k_0 e^{\alpha_i t} e^{i\beta_i t} = k_0 e^{\alpha_i t} (\cos(\beta_i t) + i\sin(\beta_i t))$$

As you can see, brackets  $(\cos(\beta_i t) + i\sin(\beta_i t))$  has a constant norm,  $||(\cos(\beta_i t) + i\sin(\beta_i t))|| = 1$ . Therefore, norm of  $z_i$  depends entirely on the norm of  $e^{\alpha_i t}$ , which is:

- constant if  $\alpha_i = 0$ , hence the system is stable.
- ② decreasing if  $\alpha_i < 0$ , hence the system is asymptotically stable.
- $\bullet$  increasing if  $\alpha_i > 0$ , hence the system is unstable.

#### General case

Consider an autonomous LTI:

$$\dot{\mathbf{x}} = \mathbf{A}\mathbf{x} \tag{12}$$

#### Definition

Eq. (12) is stable iff real parts of eigenvalues of **A** are non-positive.

#### Definition

Eq. (12) is asymptotically stable iff real parts of eigenvalues of **A** are negative.

#### Illustration

Here is an illustration of *phase portraits* of two-dimensional LTIs with different types of stability:

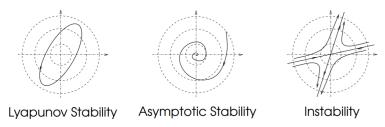


Figure 1: phase portraits for different types of stability

Credit: staff.uz.zgora.pl/wpaszke/materialy/spc/Lec13.pdf

#### READ MORE

■ Control Systems Design, by Julio H. Braslavsky staff.uz.zgora.pl/wpaszke/materialy/spc/Lec13.pdf

#### THANK YOU!

Lecture slides are available via Moodle.

You can help improve these slides at: github.com/SergeiSa/Control-Theory-Slides-Spring-2021

Check Moodle for additional links, videos, textbook suggestions.