

# 16-299 Spring 2021: Nonlinear Stability Tests

George Kantor

3 May 2021

## 1 Equilibrium Points/Stability

recall that a point  $x_e$  is an equilibrium point if  $\dot{x}$  is zero at  $x_e$ , in other words

$$f(x_e, 0) = 0$$

There is an equilibrium point at

$$x = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

which corresponds to the pendulum hanging straight down.

There is an equilibrium point at

$$x = \begin{bmatrix} \pi \\ 0 \end{bmatrix}$$

which corresponds to the pendulum standing straight up.

There are actually an infinite number of points at

$$x = \begin{bmatrix} n\pi \\ 0 \end{bmatrix}$$

for any  $n$ .

*stability:* From experience, we know that the equilibrium point at the top is unstable and the equilibrium point at the bottom is stable. If I add a little friction the equilibrium point at the bottom becomes asymptotically stable.

Let's quickly revisit our stability definitions. The equilibrium point  $x_e$  is said to be

**stable:** if for every  $\epsilon > 0$  there exists  $\delta(\epsilon) > 0$  so that if  $\|x(0)\| < \delta$  then  $\|x(t) - x_e\| < \epsilon$  for all  $t > 0$ . In words, "if you start close, you stay close".

**asymptotically stable:** if there exists an open  $D \subset \mathbb{R}^n$  with  $x_e \in D$  so that  $x(t) \rightarrow x_e$  as  $t \rightarrow \infty$  for all initial conditions  $x_0$  in  $D$ . (alternatively, we could say that there exists a  $\delta > 0$  so that if  $\|x(0)\| < \delta$  then  $x(t) \rightarrow x_e$  as  $t \rightarrow \infty$ .) In words, "if you start close, you converge."

**unstable:** if it is not stable. In words: "There are places arbitrarily close to the equilibrium point such that, if you start there, you are driven away from it."

Like before, we'd like a test to determine if the system is stable without having to integrate the nonlinear ODE. We'll learn two tests in this class, here is the first one:

## 2 Stability Test #1

### Theorem: nonlinear stability test #1

(also known as “Lyapunov’s First Method” or “Lyapunov’s Indirect Method”)

Let  $\dot{x} = f(x, u)$  be a nonlinear system with an equilibrium point at  $x_e$  and let

$$\dot{z} = Az + Bu$$

be the linearization about  $x_e$ . Let  $\Lambda = \{\lambda_1, \lambda_2, \dots, \lambda_n\}$  be the eigenvalues of  $A$ . Then

1. the system is asymptotically stable if  $Re(\lambda_i) < 0$  for all  $i = 1, 2, \dots, n$ .
2. the system is unstable if  $Re(\lambda_i) > 0$  for some  $i$ .

Note that if all of the eigenvalues are  $\leq 0$  then the test is inconclusive.

**pendulum example:** bottom eq. pt.:

$$A = \begin{bmatrix} 0 & 1 \\ -\frac{g}{\ell} & 0 \end{bmatrix}$$

and the eigenvalues of  $A$  are

$$\begin{aligned} \Lambda &= \text{roots} \left( \det \left( \begin{bmatrix} \lambda & -1 \\ \frac{g}{\ell} & \lambda + \gamma \end{bmatrix} \right) \right) \\ &= \text{roots} \left( \lambda^2 + \lambda\gamma + \frac{g}{\ell} \right) \\ \lambda &= \frac{-\gamma \pm \sqrt{\gamma^2 - 4\frac{g}{\ell}}}{2} \end{aligned}$$

Note that for positive  $g$  and  $\ell$ ,

$$\sqrt{\gamma^2 - 4\frac{g}{\ell}} < \gamma$$

so the real part of  $\lambda$  will have the opposite sign as  $\gamma$ . If  $\gamma > 0$ , then the system is stable. If  $\gamma < 0$  (i.e., negative damping) then the system is unstable.

Note that in the case where  $\gamma = 0$ , we get

$$\lambda = \pm i\sqrt{\frac{g}{\ell}}.$$

Both eigenvalues have zero real part, so test is inconclusive.

Now look at top equilibrium point:

$$A = \begin{bmatrix} 0 & 1 \\ \frac{g}{\ell} & -\text{gamma} \end{bmatrix}$$

and let’s consider the case when  $\gamma = 0$  to keep the math simple. Then

$$\begin{aligned} \Lambda &= \text{roots} \left( \det \left( \begin{bmatrix} \lambda & -1 \\ -\frac{g}{\ell} & \lambda \end{bmatrix} \right) \right) \\ &= \text{roots} \left( \lambda^2 - \frac{g}{\ell} \right) \\ &= \pm\sqrt{\frac{g}{\ell}} \end{aligned}$$

Assuming  $g > 0$  and  $\ell > 0$ , then one of these must be positive, so the system is unstable.

### 3 Stability Test #2

Recall that stability test #1 (linearize and check eigenvalues) does not always work, so we introduce a second way to test nonlinear systems for stability. This method is called “Lyapunov’s Second Method” or “Lyapunov’s Direct Method” or just “Lyapunov’s Method”. First, we assume we have an unforced nonlinear state equation with an equilibrium point at zero:

$$\dot{x} = f(x), \quad f(0) = 0.$$

**Lyapunov Function Candidate:** A Lyapunov function candidate is any differentiable function  $V : \mathbb{R}^n \rightarrow \mathbb{R}$  that satisfies

1.  $V(0) = 0$ .
2.  $V(x) > 0$  for all  $x \in D, x \neq 0$

where  $D$  is an open subset of  $\mathbb{R}^n$  that contains the point  $x = 0$ . We often call such a function a “Lyapunov candidate on  $D$ ”. Note that  $D$  can be all of  $\mathbb{R}^n$ . In plain english, near  $x = 0$ , a Lyapunov function candidate is a “bowl shaped” function with a unique minimum at  $x = 0$ .

**Time Derivative of  $V(x)$ :** We can use the chain rule to calculate how  $V$  changes with time:

$$\frac{dV(x)}{dt} = \dot{V}(x) = \frac{\partial V}{\partial x} \dot{x} = \frac{\partial V}{\partial x} f(x),$$

which can be written more explicitly as

$$\dot{V} = \begin{bmatrix} \frac{\partial V}{\partial x_1} & \frac{\partial V}{\partial x_2} & \cdots & \frac{\partial V}{\partial x_n} \end{bmatrix} \begin{bmatrix} f_1(x) \\ f_2(x) \\ \vdots \\ f_n(x) \end{bmatrix}$$

which can also be written as

$$\dot{V} = \sum_{i=1}^n \frac{\partial V}{\partial x_i} f_i(x)$$

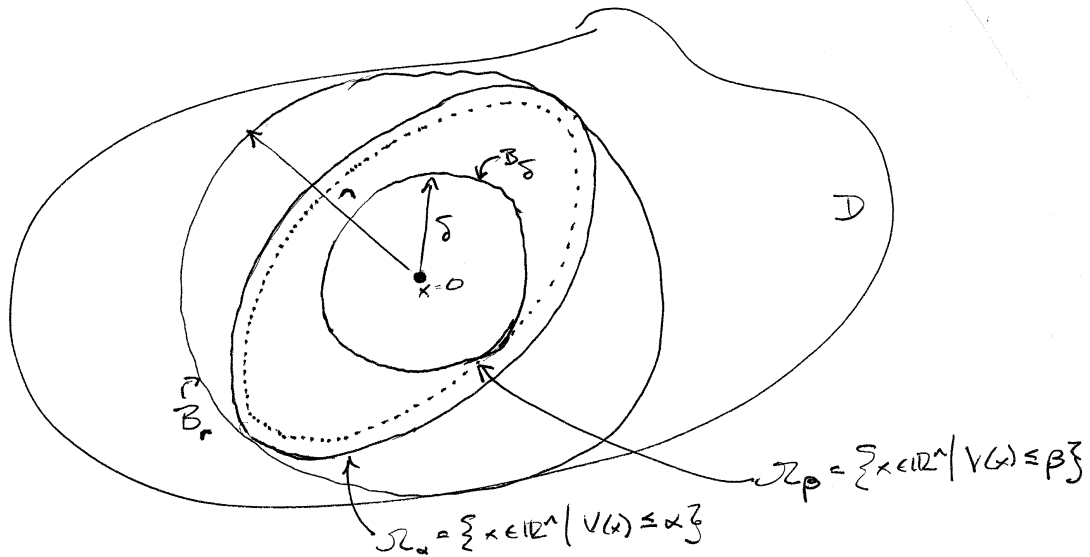
**Theorem: Stability Test #2** (aka Lyapunov’s Direct or Lyapunov’s Second Method)

Let  $V : \mathbb{R}^n \rightarrow \mathbb{R}$  be a Lyapunov function candidate on  $D \subset \mathbb{R}^n$ , where  $D$  is an open set that contains the point  $x = 0$ . The following are true:

- if  $\dot{V}(x) \leq 0$  for all  $x \in D$  then  $x = 0$  is a stable equilibrium point.
- if  $\dot{V}(x) < 0$  for all  $x \in D, x \neq 0$ , then  $x = 0$  is an asymptotically stable equilibrium point.

**Proof (stability):**

(I usually spare you the proofs, but this one is particularly intuitive, so I think it is worth going over. The math may look scary, but it keep the following picture in mind as you follow along)



Start by recalling the definition of stability: for every  $\epsilon > 0$  there exists a  $\delta > 0$  so that if  $\|x(0)\| < \delta$  then  $\|x(t)\| < \epsilon$  for all  $t \geq 0$ . In order to prove stability, we need to find a  $\delta$  and show that if  $x$  starts with  $\delta$  of zero, then it stays within  $\epsilon$  of zero (for any epsilon). Here we go:

Assume the conditions for stability from the theorem are met (i.e.,  $V$  is a Lyapunov fn. candidate,  $\dot{V} \leq 0$ )

let  $\epsilon > 0$ .

choose  $r > 0, r \leq \epsilon$  so that  $B_r \subset D$ , where

$$B_r = \{x \in \mathbb{R}^n \mid \|x\| < r\}$$

In English, choose  $r$  so that a ball of radius  $r$  is completely contained in  $D$ .

let

$$\alpha = \min_{\|x\|=r} Vx$$

we know that  $\alpha > 0$  since  $V(x) > 0$  everywhere in  $D$ .

choose  $\beta < \alpha$  and define

$$\Omega_\beta = \{x \in B_r \mid V(x) < \beta\}$$

Note that if  $x$  starts in  $\Omega_\beta$ , then it stays within  $\Omega_\beta$  for all time. (Here's why this is true: if  $x$  starts in  $\Omega_\beta$  then we know  $V(x(0)) < \beta$ . And  $\dot{V}(x) \leq 0$  so  $V$  can never grow, which means it never gets bigger than  $\beta$ , which means that  $x$  stays in  $\Omega_\beta$ .)

Now let

$$\delta = \min_{\{x \mid V(x)=\beta\}} \|x\|$$

with this choice of  $\delta$  we see that

$$\|x(0)\| < \delta \Rightarrow V(x(0)) < \beta$$

$\Rightarrow x(t)$  never leaves  $\Omega_\beta$

$\Rightarrow x(t)$  never leaves  $B_r$

$\Rightarrow \|x(t)\| < r \leq \epsilon$  for all  $t > 0$

so the equilibrium point  $x = 0$  is stable (and we're done with the proof).

**Asymptotic Stability:** the asymptotic stability part of the proof is slightly more complicated and we will not cover it in this class. The basic idea though is that when  $\dot{V}$  is strictly less than zero, then  $V$  is always decreasing, which means that  $x$  has to stay within a set that looks like the  $\Omega_\beta$  above but shrinks with time. The crux of the proof is to show that  $V(x(t)) \rightarrow 0$  instead of asymptotically converging to some nonzero value.

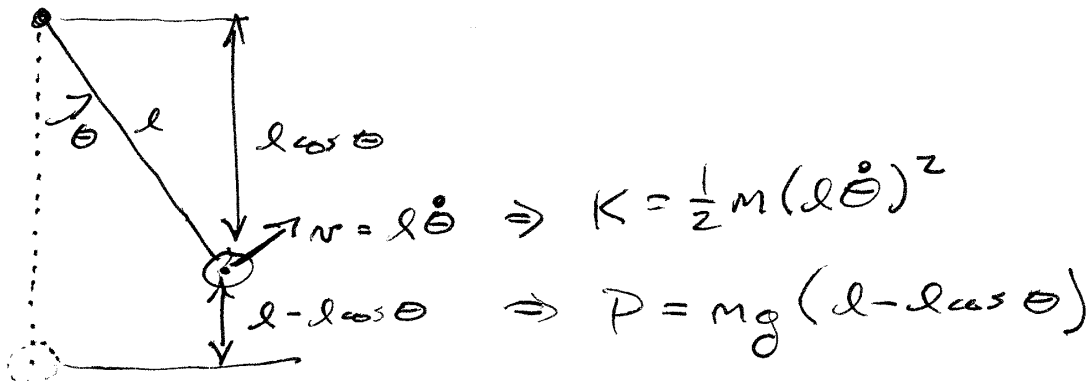
### Example: Pendulum Revisited

Recall from last time that the first stability test failed to produce a result for the case of the “straight down” equilibrium point because the eigenvalues of the linearization were both on the imaginary axis. We know intuitively that this equilibrium point should be stable, now we will try to prove it using stability test #2.

Recall that the unforced equations of motion are

$$\dot{x} = \begin{bmatrix} x_2 \\ -\frac{g}{\ell} \sin x_1 \end{bmatrix}$$

where  $x = [\theta, \dot{\theta}]^T$ .



Energy is a natural choice for a Lyapunov candidate function – we know that it is non-negative, and we can define the potential energy so that the energy at the bottom equilibrium point is zero. The kinetic energy for the pendulum is

$$K = \frac{1}{2} m (\ell \dot{\theta})^2$$

and the potential energy is

$$P = mg(\ell - \ell \cos \theta)$$

So we can define a Lyapunov function candidate:

$$V(x) = K + P = \frac{1}{2} m \ell^2 x_2^2 + mgl - mgl \cos x_1$$

Computing  $\dot{V}$ :

$$\dot{V} = \begin{bmatrix} \frac{\partial V}{\partial x_1} & \frac{\partial V}{\partial x_2} \end{bmatrix} \begin{bmatrix} f_1(x) \\ f_2(x) \end{bmatrix}$$

$$\begin{aligned}
&= [mg\ell \sin x_1 \quad m\ell^2 x_2] \begin{bmatrix} x_2 \\ -\frac{g}{\ell} \sin x_1 \end{bmatrix} \\
&= 0
\end{aligned}$$

So  $\dot{V} \leq 0$ , which means we can conclude that the bottom equilibrium point is stable!

## 4 Nonlinear Stability Summary

We have two tests for stability:

**test #1:** linearize and look at eigenvalues of resulting  $A$  matrix:

- $\text{Re}(\lambda_i) < 0$  for all  $i \Rightarrow$  asymptotically stable
- $\text{Re}(\lambda_i) > 0$  for some  $i \Rightarrow$  unstable

Note that you can never conclude stability from test #1.

**test #2:** find Lyapunov function  $V(x)$ :

- $\dot{V}(x) < 0$  for all  $x$  in a neighborhood of  $x = 0 \Rightarrow$  asymptotically stable
- $\dot{V}(x) \leq 0$  for all  $x$  in a neighborhood of  $x = 0 \Rightarrow$  stable

Note that you can never conclude instability from test #2.