Chapter 9 Global Nonlinear Techniques

Consider nonlinear dynamical system

$$X' = F(X) = \begin{pmatrix} f_1(X) \\ f_2(X) \\ \vdots \\ f_n(X) \end{pmatrix}$$

• Nullcline

- $-x_{j} nullcline = \{X : f_{j}(X) = 0\}$
- equilibrium solutions = intersection of all x_j nullclines
- $-x_j nullcline$ is usually a surface of co-dimension one
- total N x_j nullcline, j = 1, ..., n, divide the space into at least 2^n parts
- type of equilibrium can be easily figured out by looking at one direction in each part
- In $\mathbb{R}^2,$ there are two nullcline curves. They divide the plane into four parts. Each will be one of the followings
 - * NE (in which all directions in the directional field point to northeast)
 - * NW (in which all directions in the directional field point to northwest)
 - * SE (in which all directions in the directional field point to southeast)
 - * SW (in which all directions in the directional field point to southwest)

Example 1 Consider (p. 189)

$$\begin{aligned} x' &= y - x^2\\ y' &= x - 2 \end{aligned}$$

- $x nullcline : parabola y = x^2$
- y nullcline : vertical line x = 2



- there is only one equilibrium $X_0 = (2, 4)$
- the line x = 2 and the parabola $y = x^2$ divide the plane into four parts.
 - In part A containing X = (0, 4), F(X) = (4, -2) (SE)
 - In part B containing X = (3, 10), F(X) = (1, 1) (NE)
 - In part C containing X = (3,0), F(X) = (-9,1) (NW)
 - In part D containing X = (1, 0), F(X) = (-1, -1) (SW)
- X_0 is a saddle:
 - Solutions with initial values in part B and D will stay in the same regions and move away from the origin
 - Solutions with initial values in part A and C will either enter B or D, or remain in the same region. In the latter case, the solutions will converge to X_0 (stable curve). All other solutions once enter B or D, they will stay there and move away from the origin.
 - Its linearization at $X_0 = (2, 4)$ is

$$\begin{aligned} x' &= -4x + y\\ y' &= x \end{aligned}$$

- For $A = \begin{pmatrix} -4 & 1 \\ 1 & 0 \end{pmatrix}$, the eigenvalues are $\lambda = -2 \pm \sqrt{5}$. For $\lambda = -2 - \sqrt{5}$, one eigenvector is $X = (1, 2 - \sqrt{5}) = (1, -0.236)$. So

the stable curve passes through X_0 with the tangent parallel to the direction (1, -0.236)

- Recall **Stability**:
 - An equilibrium X_0 is called **stable** if any solution with initial value X_1 near X_0 will remain nearby.
 - For linear system, sinks, spiral sinks, and centers are stable.
 - An equilibrium X_0 is called **asymptotically stable** if it is stable and it converges to X_0 as $t \to \infty$.
 - Stability Theorem: Any equilibrium that is a sink or spiral sink for its linearization is asymptotically stable.
 - In other words, an equilibrium X_0 is asymptotically stable if all eigenvalues of $DF(X_0)$ have negative real parts.
- We shall extend this to the case when some eigenvalues have zero real parts.
- Liapunov Stability Analysis
 - Liapunov function for the dynamical system around X_0 : A function defined in a neighborhood O of X_0 satisfying
 - **a** $L(X_0) = 0$, L(X) > 0 for $X \neq X_0$, $X \in O$ **b** $\nabla L(X) \cdot F(X) \leq 0$ for all $X \in O$
 - Liapunov Stability Theorem: Let X_0 be an equilibrium for X = F(X). The equilibrium is stable if there exists a smooth Liapunov function. If furthermore, the Liapunov function satisfies
 - **c** $\nabla L(X) \cdot F(X) < 0$ for $X \neq X_0$ then X_0 is asymptotically stable (i.e., $X(t) \to X_0$ as $t \to \infty$)
- Note that if all eigenvalues of $DF(X_0)$ are negative, then we define

$$L(X) = -\frac{1}{2} (X - X_0)^T DF(X_0) (X - X_0) \ge 0$$

one can verify this is a Liapunov function. This is because $DF(X_0)$ is negatively definite, and, by Taylor expansion

$$F(X) = F(X_0) + DF(X_0) (X - X_0) + O(||X - X_0||^2)$$

= DF(X_0) (X - X_0) + O(||X - X_0||^2)

and

$$\nabla L(X) = -DF(X_0)(X - X_0).$$

 So

$$\nabla L(X) \cdot F(X) = -(DF(X_0)(X - X_0)) \cdot (DF(X_0)(X - X_0)) + O(||X - X_0||^3) < 0$$

as
$$X \to X_0$$
. Liapunov Theorem includes Stability Theorem.

Example 2. Consider, for parameter ε ,

$$x' = (\varepsilon x + 2y) (z + 1)$$

$$y' = (-x + \varepsilon y) (z + 1)$$

$$z' = -z^{3}$$

The only equilibrium is $X_0 = 0$, and its linearization is

$$x' = \varepsilon x + 2y$$

$$y' = -x + \varepsilon y$$

$$z' = 0$$

with eigenvalues $0, \varepsilon \pm i\sqrt{2}$. So this is not hyperbolic. Therefore, the linearization does not indicate anything about the nonlinear system. For $\varepsilon < 0$, we look for a Liapunov function in the form:

$$L\left(x, y, z\right) = ax^2 + by^2 + cz^2$$

We see that

$$\nabla L \cdot F = 2 (ax, by, cz) \cdot F$$

= $2ax (\varepsilon x + 2y) (z + 1) + 2by (-x + \varepsilon y) (z + 1) + 2cz (-z^3)$
= $2\varepsilon (ax^2 + by^2) (z + 1) + (2a - b) yx (z + 1) - 2cz^4$

So if we choose a = 1, b = 2, c = 1. Then for all X

$$\nabla L \cdot F(X) = 2\varepsilon \left(ax^2 + by^2\right) (z+1) - 2cz^4 \le 0$$

$$\nabla L \cdot F(X) < 0 \text{ if } X \ne 0$$

Hence, the equilibrium $X_0 = 0$ is asymptotically stable. However, for $\varepsilon = 0$, we can only conclude that the equilibrium is stable.

- Justification of Liapunov Theorem:
 - For any solution X(t) of the system X' = F(X(t)), we have (by (b))

$$\frac{d}{dt}L(X(t)) = \nabla L(X(t)) \cdot X'(t) = \nabla L(X(t)) \cdot F(X(t)) \le 0$$

- So L(X(t)) decreases to L(X(0))
- For any $\alpha > 0$, the set $G = \{X : L(X) < \alpha\}$ is a neighborhood of X_0 .
- For any $X_0 \in G$, along the solution X(t) initiated from X_0 , since L(X(t)) decreases to $L(X_0)$

$$L\left(X\left(t\right)\right) \le L\left(X_{0}\right) < \alpha$$

- So the entire solution X(t) remains in G — stable

Example 3. (Nonlinear pendulum) Consider a pendulum consisting of a light rod of length l to which is attached a ball of mass m. The other end of the rod is attached to a point on the ceiling. The position of the mass is described by the angle $\theta(t)$ from the straight-down position and measured in the counterclockwise direction. So the position of the mass is $l(\sin \theta(t), -\cos \theta(t))$, and velocity and acceleration are, respectively

$$v = l(\cos\theta, \sin\theta) \theta'$$

$$a = v' = l(\cos\theta, \sin\theta) \theta'' + l(-\sin\theta, \cos\theta) \theta'$$

$$F = (0, -mg) - bl(\cos\theta, \sin\theta) \theta'$$

We assume the only forces are gravitational force and Stoke's friction to be proportional to its velocity:

$$F = (0, -mg) - blv$$

= (0, -mg) - bl (cos θ , sin θ) θ' .

Now Newton's law F = ma along the tangential direction $(\cos \theta, \sin \theta)$ leads to

$$ma \cdot (\cos\theta, \sin\theta) = F \cdot (\cos\theta, \sin\theta)$$

or nonlinear Pendulum model:

$$ml\theta'' + bl\theta' + mg\sin\theta = 0$$

Set l = g = m = 1 In the system form, it is

$$\begin{aligned} \theta' &= v\\ v' &= -bv - \sin \theta \end{aligned}$$

The total energy functional can be used as its Liapunov function:

$$E\left(\theta, v\right) = \frac{1}{2}v^{2} + 1 - \cos\theta$$

One can verify that, for $b \ge 0$,

$$\nabla E(\theta, v) \cdot F = (\sin \theta, v) \cdot (v, -bv - \sin \theta) = -bv^2 \le 0$$

Thus the equilibrium $\theta = 0, v = 0$ is stable.

Example 4. Show (0,0) is asymptotically stable for

$$x' = -\frac{1}{2}x + x^{2} + 2y^{2}$$
$$y' = -x - y + 2x^{2}$$

Sol: We try $L = ax^2 + by^2$ for some a, b > 0.

$$\nabla L = (2ax, 2by)$$

$$\nabla L \cdot F = 2ax \left(-\frac{1}{2}x + x^2 + 2y^2 \right) + 2by \left(-x - y + 2x^2 \right)$$

$$= -ax^2 + 2ax^3 + 4axy^2 - 2bxy - 2by^2 + 4bx^2y$$

$$= -\left(ax^2 + 2by^2 + 2bxy \right) + 2ax^3 + 4axy^2 + 4bx^2y.$$

Choose b = 1, a = 2 we have

$$\nabla L \cdot F = -(x^2 + y^2 + 2xy) - x^2 - y^2 + 2ax^3 + 4axy^2 + 4bx^2y$$
$$= -(x + y)^2 - x^2(1 - 4x - 4y) - y^2(1 - 8x)$$

When |x| < 1/8, |y| < 1/8,

$$\nabla L \cdot F < 0.$$

Therefore, it is asymptotically stable.

- Some special nonlinear systems
 - 1. Gradient Flows: V(X) is a smooth function $\mathbb{R}^n \to \mathbb{R}^1$

$$X' = -\nabla V\left(X\right)$$

- For any solution X(t),

$$\frac{d}{dt}V(X(t)) = \nabla V(X(t)) \cdot X'(t) = -\left|\nabla V(X(t))\right|^2 \le 0$$

- So the potential function V decreases along any solution curve.
- For any level surface $\{X : V(X) = c\}$ of V, its tangent plane has the normal direction ∇V
- Therefore, any solution curve is moving towards lower-value level surfaces and is perpendicular to level surface
- Any solution X(t) will approach to a point that reaches a minimal value of V
- Critical points X_0 (i.e., $\nabla V(X_0) = 0$) of V are equilibrium solutions.
- All equilibria are stable.
- Any isolated local minimum points X_0 are asymptotically stable.
- Linearization matrices are symmetric, and have only real eigenvalues.

Example 5 Consider gradient system

$$V = x^2 (x - 1)^2 + y^2$$



Since

$$\nabla V = \left(2x\left(2x^2 - 3x + 1\right), \ 2y\right)$$

there are three critical points: $X_0 = (0,0), (1,0), (0.5,0)$. The first two are local minimums of V. But (0.5,0) is not.

2. Hamiltonian Systems: H(X) is smooth function (called Hamiltonian function)

$$x' = \frac{\partial H}{\partial y}$$
$$y' = -\frac{\partial H}{\partial x}$$

-H(X) is constant along any solution X(t)

$$\frac{d}{dt}H(X(t)) = \nabla H(X(t)) \cdot X'(t) = (\partial_x H, \partial_y H) \cdot (\partial_y H, -\partial_x H) = 0$$
$$H(X(t)) = H(X(0))$$

- thus, Hamiltonian system is conservative system: it will not alter value of H
- Therefore, a solution curve is a part of a level curve of H
- linearizations have the structure

$$A = \begin{pmatrix} \frac{\partial^2 H}{\partial x \partial y} & \frac{\partial^2 H}{\partial y^2} \\ -\frac{\partial^2 H}{\partial x^2} & -\frac{\partial^2 H}{\partial x \partial y} \end{pmatrix}$$

- Its characteristic polynomial is

$$\lambda^2 - \left(\frac{\partial^2 H}{\partial x \partial y}\right)^2 + \frac{\partial^2 H}{\partial x^2}\frac{\partial^2 H}{\partial y^2} = 0$$

- Eigenvalues of any linearized Hamiltonian system at critical point $\nabla H(X_0) = 0$:
 - * If det $(D^2H) = \frac{\partial^2 H}{\partial x^2} \frac{\partial^2 H}{\partial y^2} \left(\frac{\partial^2 H}{\partial x \partial y}\right)^2 > 0$ (local minimum of H), then eigenvalue are $\pm \sqrt{\det(D^2H)}i$, and thus X_0 is a center.
 - * If det $(D^2H) < 0$ (local saddle of H), then eigenvalue are $\pm \sqrt{-\det(D^2H)}$, and thus X_0 is a saddle.

Example 6. (ideal pendulum) The frictionless pendulum

$$\begin{aligned} \theta' &= v\\ v' &= -\sin\theta \end{aligned}$$

is a Hamiltonian with

$$H\left(\theta,v\right) = \frac{1}{2}v^{2} + 1 - \cos\theta$$

Example 7. Consider

$$\begin{aligned} x' &= y\\ y' &= -x^3 + x \end{aligned}$$

This is a Hamiltonian with

$$H = \frac{x^4}{4} - \frac{x^2}{2} + \frac{y^2}{2}$$

There are three equilibrium solutions $X_0 = (0,0), (\pm 1,0)$. Note that the linearized system is X' = AX

$$A = \left(\begin{array}{cc} 0 & 1\\ 1 - 3x^2 & 0 \end{array}\right)$$

while the Hessian matrix is

$$D^2H = \left(\begin{array}{cc} 3x^2 - 1 & 0\\ 0 & 1 \end{array}\right)$$

At (0,0), D^2H is neither positive nor negative, so (0,0) is a saddle for H, or a saddle for the linearized system X' = A(0,0) X. At the other two equilibria $X_0 = (\pm 1,0)$, D^2H is positively definite. So they are local minimum points of H, and are center for the linearized systems.

- Homework: 1ab, 2, 6, 7aef (no phase portrait, no level surface)
- #8 (no phase portrait. Do the followings)
 - 1. Determine which is gradient, which is Hamiltonian.
 - 2. (optional) If it is a gradient or Hamiltonian, find V or H.
 - Hints for #8: In 2d, a system

$$\begin{aligned} x' &= f \\ y' &= g \end{aligned}$$

a is a gradient iff

$$\frac{\partial f}{\partial y} = \frac{\partial g}{\partial x}$$

then, V can be found by solving

$$\frac{\partial V}{\partial x} = f$$
$$\frac{\partial V}{\partial y} = g$$

b is a Hamiltonian iff

$$\frac{\partial f}{\partial x} = -\frac{\partial g}{\partial y}$$