

Chapter 7 Nonlinear Systems

Nonlinear systems in R^n :

$$X' = F(t, X)$$

$$X = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}, F(t, X) = \begin{pmatrix} F_1(t, x_1, \dots, x_n) \\ \vdots \\ F_n(t, x_1, \dots, x_n) \end{pmatrix}$$

When $F(t, X) = F(X)$ is independent of t , it is an example of dynamical system.

- Dynamical System:

- Definition: Consider any smooth two-variable function $\phi(t, X) : (t, X) \in R^1 \times R^n \mapsto R^n$. Let $\phi_t(X) = \phi(t, X)$ be a map $R^n \mapsto R^n$. We call the family of map $\{\phi_t\}$ a smooth dynamical system if it satisfies the following two conditions

1. ϕ_0 is an identity map: $\phi_0(X) = \phi(0, X) = X$ for any X
2. $\phi_t \circ \phi_s = \phi_{t+s} : \phi(t, \phi(s, X)) = \phi(t+s, X)$

- Flows of linear systems $X' = AX$ are dynamical systems

- * Solution is

$$\phi_t(X_0) = \phi(t, X_0) = e^{tA}X_0$$

- * So

$$\phi_0(X_0) = e^{0A}X_0 = X_0$$

$$\phi_t \circ \phi_s(X_0) = \phi_t(e^{sA}X_0) = e^{tA}e^{sA}X_0 = e^{(t+s)A}X_0 = \phi_{t+s}(X_0)$$

- Flows of nonlinear systems $X' = F(X)$ are also dynamical systems (if F is smooth function)

- Integral equation formulation:

$$X(t) = X_0 + \int_0^t F(X(s)) ds$$

- Picard Iteration: To solve this system, we create an iterative sequence

$$X_0 = X_0 \quad (\text{initial value})$$

$$X_{n+1} = X_0 + \int_0^t F(X_n(s)) ds$$

- Existence (for general system $F(t, X)$):
 - * the Picard sequence $\{X_n(t)\}$ has a limit as $n \rightarrow \infty$, for a small time period $-\varepsilon < t < \varepsilon$. (local existence theorem)
 - * the limit $X(t)$ is a solution for t in $(-\varepsilon, \varepsilon)$, because it satisfies the integral formulation
 - * For linear (nonautonomous system)

$$X' = A(t)X$$

the Picard sequence converges for all t as long as $A(t)$ is defined and continuous. (Global existence)

- Uniqueness (for general system $F(t, X)$): The solution is unique, i.e., there is only one solution for the same IVP.
- Continuous dependence on initial data (for general system $F(t, X)$):
 - * For Y_0 close to X_0 , solution $Y(t)$ with the initial data Y_0 also exists in $(-\varepsilon, \varepsilon)$
 - * From integral formulations

$$Y(t) = Y_0 + \int_0^t F(Y(s)) ds$$

we see

$$X(t) - Y(t) = X_0 + \int_0^t F(X(s)) ds - \left(Y_0 + \int_0^t F(Y(s)) ds \right)$$

So if $|\nabla F| \leq K$,

$$\begin{aligned} |X(t) - Y(t)| &\leq |X_0 - Y_0| + \int_0^t |F(X(s)) - F(Y(s))| ds \\ &\leq |X_0 - Y_0| + \int_0^t |F(X(s)) - F(Y(s))| ds \\ &\leq |X_0 - Y_0| + K \int_0^t |X(s) - Y(s)| ds \end{aligned}$$

* By Gronwall's inequality, we arrive at

$$|X(t) - Y(t)| \leq |X_0 - Y_0| e^{Kt}$$

* So if $Y_0 \rightarrow X_0$, then $Y(t) \rightarrow X(t)$

– Continuous dependence on parameters (for general system $F(t, X)$):

$$X' = F_a(X)$$

* Let $X_a(t)$ be the solution, and $X_b(t)$ the solution for $X' = F_b(X)$. Then

$$X_a(t) - X_b(t) = \int_0^t (F_a(X_a(s)) - F_b(X_b(s))) ds$$

* So if $F_b(X) \rightarrow F_a(X)$ as $b \rightarrow a$, $X_b(t) \rightarrow X_a(t)$

– So the flow $\phi(t, X)$ exists and unique, and continuously depends on initial data X .

– The flow is a dynamical system:

* By definition, $\phi(0, X) = X$

* For any fixed s , set $Y(t) = \phi(t + s, X)$. Then $Y(t)$ solves

$$Y'(t) = \frac{\partial}{\partial t} \phi(t + s, X) = F(\phi(t + s, X)) = F(Y(t))$$

$$Y(0) = \phi(s, X)$$

* Recall that $W(t) = \phi(t, X_0)$ solves

$$Y'(t) = F(Y(t))$$

$$Y(0) = X_0$$

* In particular, for $X_0 = \phi(s, X)$, $W(t) = \phi(t, \phi(s, X))$ solves

$$Y'(t) = F(Y(t))$$

$$Y(0) = \phi(s, X)$$

* By the uniqueness theory, $W(t) = Y(t)$

$$\phi(t, \phi(s, X)) = \phi(t + s, X)$$

– Integral formulation for flow

$$\phi_t(X) = X + \int_0^t F(\phi_s(X)) ds$$

• All dynamical systems are flows of ODEs:

– Given $\phi_t(X)$, we compute $F(X)$

$$\frac{\partial}{\partial t} \phi_t(X) |_{t=0} = F(X)$$

– Since $\phi_{t+s}(X) = \phi_t(\phi_s(X))$

$$\frac{\partial}{\partial t} \phi_{t+s}(X) = \frac{\partial}{\partial t} (\phi_t(\phi_s(X)))$$

– At $t = 0$,

$$\frac{\partial}{\partial t} \phi_{t+s}(X) = \frac{\partial}{\partial s} \phi_s(X)$$

$$\frac{\partial}{\partial t} (\phi_t(\phi_s(X))) = F(\phi_s(X))$$

– so $\phi_s(X)$ is a solution of $Y' = F(Y)$

Example 1: Use the Picard iteration to solve

$$X' = \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix} X, \quad X(0) = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

Sol: Recall that the solution is

$$X = \begin{pmatrix} \cos t \\ -\sin t \end{pmatrix}$$

We now generate the Picard sequence:

$$U_{n+1} = \begin{pmatrix} 1 \\ 0 \end{pmatrix} + \int_0^t \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix} U_n(s) ds$$

and show it converges to X .

- Jacobian matrix of mappings $F : R^n \rightarrow R^n$

$$F(X) = \begin{pmatrix} f_1(x_1, \dots, x_n) \\ \dots \\ f_n(x_1, \dots, x_n) \end{pmatrix}, \quad DF(X) = \begin{pmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} & \dots & \frac{\partial f_1}{\partial x_n} \\ \dots & \dots & \dots & \dots \\ \frac{\partial f_n}{\partial x_1} & \frac{\partial f_n}{\partial x_2} & \dots & \frac{\partial f_n}{\partial x_n} \end{pmatrix} = \begin{pmatrix} \nabla f_1 \\ \dots \\ \nabla f_n \end{pmatrix}$$

Example 2 (a) For $F(X) = X$ in R^n , $DF(X) = I$

(b) In R^2 , $F = (x_1x_2, x_1^2)$

$$DF = \begin{pmatrix} x_2 & x_1 \\ 2x_1 & 0 \end{pmatrix}$$

- Some formulas: $DF(\phi) = DF \cdot D\phi$

- Differentiate DE

$$\frac{\partial \phi(t, X)}{\partial t} = F(\phi(t, X))$$

with respect to X , we see $D\phi(t, X) = D\phi_t(X)$ Solves

$$\frac{\partial D\phi(t, X)}{\partial t} = DF(\phi(t, X)) D\phi(t, X)$$

i.e., each column of $D\phi(t, X)$ solves

$$U' = DF(\phi(t, X))U$$

or

$$D\phi_t(X) = I + \int_0^t DF(\phi_s(X)) D\phi_s(X) ds$$

- The Variational Equation

- Given X_0 , let $\phi(t, X_0)$ be the solution of IVP with the initial value X_0
- Set $A(t) = DF(\phi(t, X_0))$, i.e., the Jacobian matrix of F evaluated at $\phi(t, X_0)$
- We call the following **Variational Equation** along solution $\phi(t, X_0)$:

$$U' = A(t)U$$

- Variational equation is a (global) linearization along $\phi(t, X_0)$:
 - If $\phi(t, X_0)$ exists for t in $[0, T]$, then so does the entire flow $\psi(t, U_0)$ for Variational equation, i.e.,

$$\frac{\partial \psi(t, U_0)}{\partial t} = A(t) \psi(t, U_0)$$

- Since

$$\frac{\partial D\phi(t, X_0)}{\partial t} = DF(\phi(t, X_0)) D\phi(t, X_0)$$

- each column of $D\phi(t, X_0)$ is a solution of the variational equation
- Multiplying both sides by U_0 :

$$\frac{\partial D\phi(t, X_0) U_0}{\partial t} = DF(\phi(t, X_0)) D\phi(t, X_0) U_0$$

- Since $D\phi(0, X_0) U_0 = U_0$,

$$\psi(t, U_0) = D\phi(t, X_0) U_0$$

- This means that, if we are able to solve the viational equation, then we can easily find $D\phi(t, X_0)$ since

$$D\phi(t, X_0) = [\psi(t, e_1), \psi(t, e_2), \dots, \psi(t, e_n)]$$

where e_i is the standard basis.

- If $\phi(t, X)$ is smooth, we expand $\phi(t, X_1)$ in X variable around X_0 :

$$\phi(t, X_1) = \phi(t, X_0) + D\phi(t, X_0) (X_1 - X_0) + O(|X_1 - X_0|^2)$$

- So if $X_1 = X_0 + U_0$, then

$$\phi(t, X_1) = \phi(t, X_0) + \psi(t, U_0) + O(|X_1 - X_0|^2)$$

This means that if we can solve IVP at initial value X_0 to get $\phi(t, X_0)$, then we can calculate $A(t) = DF(\phi(t, X_0))$ and thus to calculate the solution $\psi(t, U_0)$ for the viational equation since it is linear. Then, we can approximate other IVP for nearby initial value $X = X_0 + U_0$ by

$$\phi(t, X_0) + \psi(t, U_0)$$

- In general (with less smoothness condition), let $X(t) = \phi(t, X_0)$, $U(t) = \psi(t, U_0)$, then $X(t) + U(t)$ is an approximation for $\phi(t, X_0 + U_0)$ of order higher than one:

$$\frac{|X(t) + U(t) - \phi(t, X_0 + U_0)|}{|U_0|} \rightarrow 0$$

as $|U_0| \rightarrow 0$. The convergence is uniform in t .

- In particular, if X_0 is an equilibrium solution, i.e., $\phi(t, X_0) = X_0$, then

$$A(t) = DF(\phi(t, X_0)) = DF(X_0)$$

Variational equation is

$$U' = DF(X_0)U$$

Its solution $\psi(t, U_0)$ is such that

$$\phi(t, X_0) + \psi(t, U_0) = X_0 + \psi(t, U_0)$$

is an approximation solution for $\phi(t, X_0 + U_0)$. So sometime we call the variational equation linearization.

Example 3 $x' = x^2$. Solution is

$$\phi(t, x_0) = \frac{x_0}{1 - x_0 t}$$

So

$$D\phi(t, x_0) = \frac{1}{(1 - x_0 t)^2}$$

On the other hand, $DF = 2x$. So its variational equation for $x = \phi(t, x_0)$ is

$$u' = 2\phi(t, x_0)u = \left(\frac{2x_0}{1 - x_0 t}\right)u$$

The solution of this variational equation with initial data u_0 is

$$\psi(t, u_0) = \frac{u_0}{(1 - x_0 t)^2}$$

Apparently, this function is defined as long as $t < 1/x_0$, the same as for $\phi(t, x_0)$.

For $x_1 = x_0 + u_0$,

$$\begin{aligned}
 & \phi(t, x_0) + \psi(t, u_0) - \phi(t, x_1) \\
 &= \frac{x_0}{1 - x_0 t} + \frac{u_0}{(1 - x_0 t)^2} - \frac{x_1}{1 - x_1 t} \\
 &= \frac{x_1 - x_0^2 t}{(1 - x_0 t)^2} - \frac{x_1}{1 - x_1 t} \\
 &= \frac{x_1 - x_0^2 t - x_1^2 t + x_0^2 x_1 t^2 - x_1 (1 - x_0 t)^2}{(1 - x_0 t)^2 (1 - x_1 t)} \\
 &= \frac{-(u_0)^2 t}{(1 - x_0 t)^2 (1 - x_1 t)}
 \end{aligned}$$

So as $u_0 \rightarrow 0$,

$$\frac{|\phi(t, x_0) + \psi(t, u_0) - \phi(t, x_1)|}{|u_0|} = \left| \frac{u_0 t}{(1 - x_0 t)^2 (1 - x_1 t)} \right| \rightarrow 0$$

- When X_0 is an equilibrium solution, the variational equation becomes an autonomous linear system. We call in this case the linearization.

Example 5: Consider $X' = F(X)$, $F = (x + y^2, -y)$. $X = 0$ is an equilibrium. So along this solution, the linearization is

$$DF(0) = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}, \quad U(t) = \begin{pmatrix} x_0 e^t \\ y_0 e^{-t} \end{pmatrix}$$

- Homework: 1ce, 2, 7(for the case (i) $A(t) = \text{diag}(\lambda_1(t), \dots, \lambda(t))$, (ii) $n = 2$, general matrices that may not be diagonal)
- Homework (additional): Find and solve the variational equations for $X' = F(X)$
 1. $F = (x^2 + xy, x + y^3)$, $X = (-1, 1)$
 2. $x' = x^{4/3}$, $x(t) = 27(3 - t)^{-3}$
- Homework (optional): Write (or find from internet) a numerical program (in any language, such as C, Matlab, Mathematica, etc.) for (i) Euler's method (ii) Runge-Kutta of order 4.