Engineering Tripos Part II

Module 4F2 -- NL Systems and control

Nonlinear systems (7 lectures)

Handout 1

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available on RS homepage



•The transfer function view: sinusoids map to sinusoids



• The state-space view: state parametrizes memory

• The interconnection view: systems are made of simpler systems

















The essence of the feedback amplifier.

The essence of control theory.





Oscillations arise from hysteresis loops : a combination of positive and negative feedback

An essential nonlinear phenomenon



An essential source of nonlinearity.

'Angular' spaces: circle, rotation group, sphere, ...



Linear versus nonlinear analysis

In linear systems analysis, the emphasis in on the solution at time t, or the flow:

$$\phi(t, x_0) = e^{At} x_0$$

In nonlinear systems analysis, the flow can only be approximated, typically with the help of a numerical integrator.

Therefore, the emphasis is on the vector field and trajectories in the state space.

The central question is : what can be said about the asymptotic behavior (large times) without integrating the vector field?

A problem of historical importance

Newton proposes a general law for motions: F=m a (1689). Solves the two-body problem and 'proves' Kepler laws for the behavior 'earth+sun'.

(Kepler laws are about integral curves. Newton law is about the vector field). Newton also invents a calculus to approximately solve nonlinear differential equations.

Laplace raises the question of the asymptotic behavior: Is the solar system stable? Will the observed behavior persist eternally?

Poincaré puts an end to the attempt to answer asymptotic questions from approximate solutions for the three-body problem. He invents the geometric analysis of nonlinear systems, based on a study of the vector field in the state space.

Vector fields on the line

$$\dot{x} = \sin(x) + u$$

- x is a real number, denoting position on the real line.
- $\dot{x}\,$ is a real number, the sign of which indicates the direction of motion





$$\dot{\theta} = \sin(\theta) + u$$

- θ is an angle, denoting position on the circle.
- $\dot{\theta}$ is a real number, the sign of which indicates the direction of motion in the tangent space $T_{\theta}S^1 \approx \mathbb{R}$



Both the angle and the velocity can be represented by real numbers locally (a coordinate representation) but they are different objects!

Scalar vector fields derive from a potential

$$\dot{x} = x - x^{3}$$

$$= -\frac{\partial V}{\partial x}$$

$$V(x) = \frac{x^{4}}{4} - \frac{x^{2}}{2}$$

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By definition, the potential decreases along solutions:

$$\dot{V} = -(\frac{\partial V}{\partial x})^2 \le 0$$

This means that solutions move 'downhill' in the potential landscape.

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Existence and uniqueness of solutions

 $\dot{x} = x^{1/3}$ several solutions with initial condition

 $\dot{x}=1+x^2$ $\,\,$ solution blows up to infinity in finite time $\,$

Existence and Uniqueness Theorem: Consider the initial value problem

$$\dot{x} = f(x), \qquad x(0) = x_0.$$

Suppose that f(x) and f'(x) are continuous on an open interval R of the x-axis, and suppose that x_0 is a point in R. Then the initial value problem has a solution x(t) on some time interval $(-\tau, \tau)$ about t = 0, and the solution is unique.

Vector fields versus maps

Discrete-time and continuous-time linear systems are treated on the same foot because of the analogy between the solution of $\dot{x} = Ax$ and the solution of $x_{+} = Ax$

The analogy does not extend to nonlinear behaviors.

Trajectories of $\dot{x} = f(x)$ are integral curves of the vector field.

Trajectories of $x_+ = F(x)$ are a sequence of points generated by iterating the map

$$x_0, x_1 = F(x_0), x_2 = F(F(x_0)), \dots, x_N = F^N(x_0), \dots$$

This course is primarily about continuous-time nonlinear behaviors.

4F2: Phase portraits, linearization, and saddle points

Past lecture: static analysis and one-dimensional state-spaces.

Today: two-dimensional state-spaces.

Vector fields and phase portrait

The vector equation

$$\dot{x} = f(x), \quad x = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \in \mathbb{R}^2$$

is equivalent to two coupled scalar equations:

$$\dot{x}_1 = f_1(x_1, x_2)$$

 $\dot{x}_2 = f_2(x_1, x_2)$

Drawing the phase portrait means attaching an arrow to each point and sketching the integral curves of the vector field.



(Matlab command 'quiver' draws f(x))

Goal of lecture: understanding the behavior of two-dimensional systems from their phase portrait



Imagine a few possible solutions of this phase portrait. Where is the time information? What are the asymptotic behaviors?

Observe: trajectories do not intersect !

Isoclines: 'loci of same slope' $\begin{array}{c} x_2 \\ & &$

Linear phase portraits

$$\dot{x} = Ax \qquad \equiv \qquad \left(\begin{array}{c} \dot{x}_1 \\ \dot{x}_2 \end{array} \right) = \left(\begin{array}{c} a_{11} & a_{12} \\ a_{21} & a_{22} \end{array} \right) \left(\begin{array}{c} x_1 \\ x_2 \end{array} \right)$$

Solution:

$$\left(\begin{array}{c} x_1(t) \\ x_2(t) \end{array}\right) = e^{At} \left(\begin{array}{c} x_1(0) \\ x_2(0) \end{array}\right)$$

Geometry of solution is determined by the eigenvalues and eigenvectors of A

Robust phase portraits:			
Stable node	Unstable node	Saddle point	
Fragile phase portraits:			
repeated eigenvalue	zero eigenvalu		

Linear phase portraits with complex eigenvalues

Robust phase portraits:

Stable focus

Unstable focus

Fragile phase portraits:



Linearization

Consider a solution $x^*(t)$ of $\dot{x} = f(x)$

The (Jacobian) linearization (or variational equation) of $\ \dot{x}=f(x)$ along $\ x^*(t)$ is the linear system

$$\dot{z} = A(t)z, \ A(t) = \frac{\partial f}{\partial x}(x^*(t)), \ a_{ij}(t) = \frac{\partial f_i}{\partial x_j}(x^*(t))$$

The variational equation is obtained by retaining the first-order terms in the Taylor expansion of $\,\dot{x}=f(x)$ along $\,x^*(t)\,$

Linearization at hyperbolic fixed points

A fixed point is called hyperbolic if the eigenvalues of the linearization lie off the imaginary axis (nodes, foci, saddles).

Robust linear phase portraits are important because they capture the local behavior of nonlinear phase portraits near hyperbolic fixed points.



(Hartman Grobam theorem)





Reduced modeling

Compare the following three behaviors:

The static model $0 = \frac{u - x}{R} - f(x)$ The one-dimensional model $C\dot{x} = v - f(x)$ $v = \frac{u - x}{R}$ The two-dimensional model $\begin{cases} C \cdot \dot{x_1} = x_2 - f(x_1) \\ L \cdot \dot{x_2} = -R \cdot x_2 - x_1 + u \end{cases}$









Newton's law:

$$\ddot{\theta} = -\frac{g}{l}\sin(\theta)$$

State-space model:

$$\begin{cases} \dot{\theta} &= v \\ \dot{v} &= -\frac{g}{l}\sin\theta \end{cases}, \ (\theta, v) \in S^1 \times \mathbb{R}$$

The state-space is a cylinder, not a plane !





The pendulum: energy conservation



$$E = \frac{1}{2}v^2 - \frac{g}{l}\cos\theta$$

Energy conservation:

$$\dot{E} = \frac{\partial E}{\partial v}\dot{v} + \frac{\partial E}{\partial \theta}\dot{\theta} = 0$$

Vector field is everywhere tangent to the level curves of E. This means that the level curves of E are the integral curves of the vector field !





Summary of lecture

Phase portrait: two-dimensional behaviors can be drawed.

Linear phase portraits determine local behavior near hyperbolic fixed points

Tunnel diode: archetype example of bistable behavior

Pendulum: archetype example of conservative mechanical behavior.

The saddle point is an important 'hidden' fixed point. A key ruler of nonlinear behaviors.