

In this class we will model dynamic systems using state space models of the form

$$\dot{x}_1 = f_1(t, x_1, x_2, \dots, x_n, u_1, \dots, u_p)$$

$$\vdots$$

$$\dot{x}_n = f_n(t, x_1, x_2, \dots, x_n, u_1, \dots, u_p)$$

$$y_1 = h_1(t, x_1, \dots, x_n, u_1, \dots, u_p)$$

$$\vdots$$

$$y_q = h_q(t, x_1, \dots, x_n, u_1, \dots, u_p)$$

Letting $x = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}$ $u = \begin{pmatrix} u_1 \\ \vdots \\ u_p \end{pmatrix}$ $y = \begin{pmatrix} y_1 \\ \vdots \\ y_q \end{pmatrix}$

we can write

$$\dot{x} = f(t, x, u)$$

$$y = h(t, x, u)$$

The first part of the book will deal with unforced equations of the form

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

$$y = h(t, x)$$

In the time-invariant case we get

$$\dot{x} = f(x)$$

$$y = h(x)$$

which has traditionally been called an "autonomous" system.

Def: A point ~~x^*~~ x^* is called an equilibrium point if $f(x^*) = 0$, ~~which~~

Equilibrium points have the property that if $x_0 = x^*$ then $x(t) = x^* \quad \forall t \geq 0$

Note that $\dot{x} = f(x^*) = 0$ implies that x never changes.

Compare $\dot{x} = f(t, x)$
 $y = h(t, x)$

with the linear state-space model

$$\dot{x} = A(t)x + B(t)u$$

$$y = C(t)x + D(t)u$$

Recall that Laplace + Fourier transforms are useful because ~~the~~ convolution in time domain transforms to multiplication in frequency domain, and convolution completely characterizes input-output behavior for LTI systems.

For nonlinear systems, convolution no longer characterizes input-output behavior \therefore transform methods are much less useful. Almost all analysis is carried out in time-domain

Nonlinear systems have much richer behavior than linear systems.

Some "nonlinear phenomena" that do not occur in linear systems include

- Finite escape time
- Multiple isolated equilibria
- Limit cycles
- Chaos
- Bifurcation

Chap 1 intro discusses a number of examples of nonlinear systems. These examples will be used for illustration throughout the ~~book~~ book.

Example: Pendulum



Newton's 2nd law

$$J\ddot{\theta} = \sum \text{torques}$$

where $J = \int r^2 dm = ml^2$

$$ml\ddot{\theta} = \sum \text{torques}$$

torque due to gravity $\tau_g = r \times F$

$$\begin{aligned} &= |r||F|\sin\theta \\ &= lmg\sin\theta \end{aligned}$$

adding viscous friction

$$\tau_v = -k l \dot{\theta}$$

opposing force proportional to velocity

we get

$$ml\ddot{\theta} = -mgl\sin\theta - k l \dot{\theta}$$

If we apply a torque T at the hub

then the equations of motion are

$$ml\ddot{\theta} = -mgl\sin\theta - k l \dot{\theta} + T$$

Setting $\dot{x}_1 = 0$, $\dot{x}_2 = 0$, $u = T$, we get

$$\left. \begin{aligned} \dot{x}_1 &= \dot{x}_2 \\ \dot{x}_2 &= -\frac{g}{l} \sin x_1 - \frac{k}{ml} x_2 + \frac{1}{me} u \end{aligned} \right\} f(x, u)$$

The unforced equilibrium points are found by setting $f(x, 0) = 0$

ie

$$x_2 = 0$$

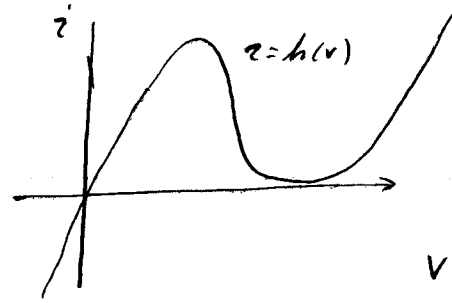
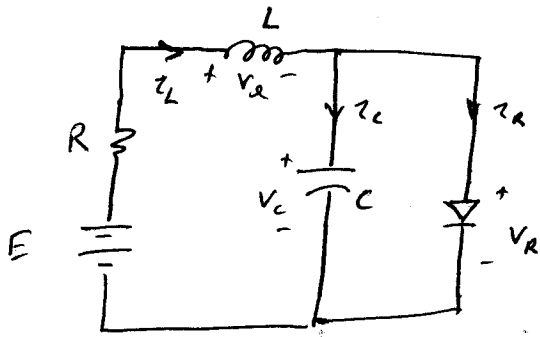
$$-\frac{g}{l} \sin x_1 - \frac{k}{ml} x_2 = 0$$

The equilibrium points are therefore $x = \begin{pmatrix} n\pi \\ 0 \end{pmatrix}$

$$n = 0, \pm 1, \dots$$

Example

Tunnel Diode Circuit



diode characteristics

$$i_c = C \frac{dv_c}{dt}$$

$$v_L = L \frac{di_L}{dt}$$

Kirchhoff's current law

$$i_c + i_R - i_L = 0$$

Kirchhoff's voltage Law

$$v_c - E + R i_L + v_L = 0$$

Define the state variables as

$$x_1 = v_c, \quad x_2 = i_L, \quad u = E$$

which gives

$$x_1 - u + R x_2 + L \frac{dx_2}{dt} = 0$$

$$C \frac{dx_1}{dt} + h(x_1) - x_2 = 0$$

or

$$\left. \begin{aligned} \dot{x}_1 &= \frac{1}{C} x_2 - \frac{1}{C} h(x_1) \\ \dot{x}_2 &= -\frac{1}{L} x_1 - \frac{R}{L} x_2 + \frac{1}{L} u \end{aligned} \right\} \dot{x} = f(x, u)$$

The ^{non-trivial} equilibrium points are found by setting

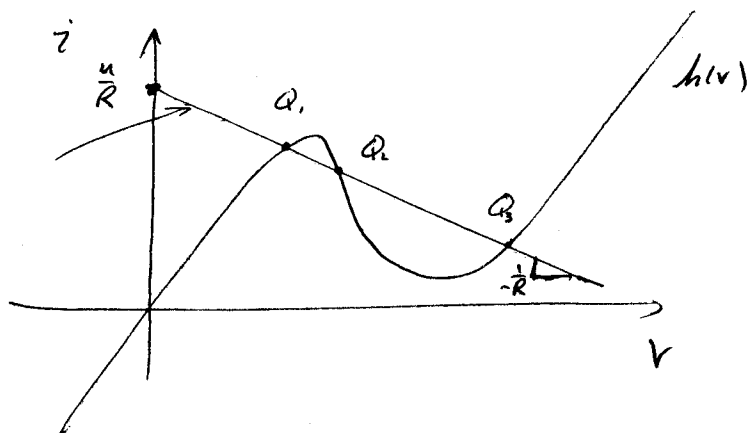
$$f(x, 0) = 0$$

ie $x_2 = h(x_1)$

$$x_1 + R x_2 = u$$

or $x_1 + R h(x_1) = u$

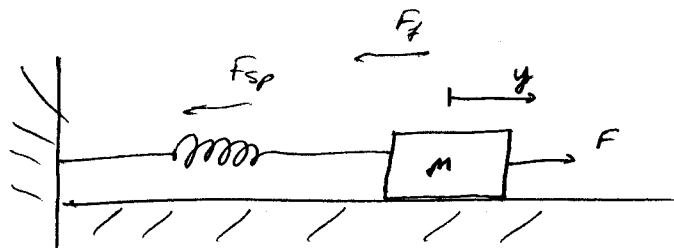
or $h(x_1) = \frac{1}{R} u - \frac{1}{R} x_1$



Notice that for various values of u , the system may have 1, 2, or 3 equilibrium points.

\therefore The very structure of the system changes.

Changes from one equilibrium point to three is called a bifurcation.

Example Mass - Spring System

$y \equiv$ position from zero spring force

$F_{sp} \equiv$ spring force

$F_f \equiv$ force due to friction

$F \equiv$ applied force

From Newton's 2nd law

$$m\ddot{y} = F - F_{sp} - F_f$$

Assume spring force is only due to displacement

ie $F_{sp} = g(y)$

A linear spring is described by Hooke's law as

$$g(y) = k_{sp} y$$

but springs are rarely linear. Common models that describe 2nd order effects are

softening springs:

$$g(y) = k(1 - a^2 y^2) y \quad (ay) < 1$$

hardening spring:

$$g(y) = ~~k(1 + a^2 y^2)~~ k(1 + a^2 y^2) y$$

There are several friction effects:

- 1) Viscous friction:
 - due to air (or fluid) resistance
 - proportional to velocity, i.e. $F_v = c\dot{y}$
- 2) Kinetic friction:
 - due to surface tension
 - $F_k = \mu_k m\dot{y}$ opposite motion
- 3) static friction:
 - due to surface tension
 - F_s acts parallel to surface + is bounded by $\mu_s m\dot{y}$ $0 < \mu_s < 1$

Kinetic + static friction can be combined to produce dry friction

$$F_d = \begin{cases} -\mu_k m\dot{y} & \text{if } \dot{y} < 0 \\ F_s & \dot{y} = 0 \\ \mu_k m\dot{y} & \text{if } \dot{y} > 0 \end{cases}$$

If the spring force is linear we have

$$m\ddot{y} + k y + c\dot{y} + \eta(y, \dot{y}) = 0$$

where

$$\eta(y, \dot{y}) = \begin{cases} \mu_k m\dot{y} \operatorname{sign}(\dot{y}) & |\dot{y}| > 0 \\ -k y & \dot{y} = 0 \quad |\dot{y}| \leq \mu_s m\dot{y}/k \\ -\mu_s m\dot{y} \operatorname{sign}(\dot{y}) & \dot{y} = 0 \quad |\dot{y}| \geq \mu_s m\dot{y}/k \end{cases}$$

Letting $x_1 = y$, $x_2 = \dot{y}$ the state space model is

$$\dot{x}_1 = x_2$$

$$\dot{x}_2 = -\frac{k}{m} x_1 - \frac{c}{m} x_2 - \frac{1}{m} \eta(x_1, x_2)$$

Note: Many equilibria, $f(x)$ is discontinuous.

However note that when

$x_2 > 0$ we can write

$$\dot{x}_1 = x_2$$

$$\dot{x}_2 = -\frac{k}{m} x_1 - \frac{c}{m} x_2 - M_0 g$$

and when $x_2 < 0$ we can write

$$\dot{x}_1 = x_2$$

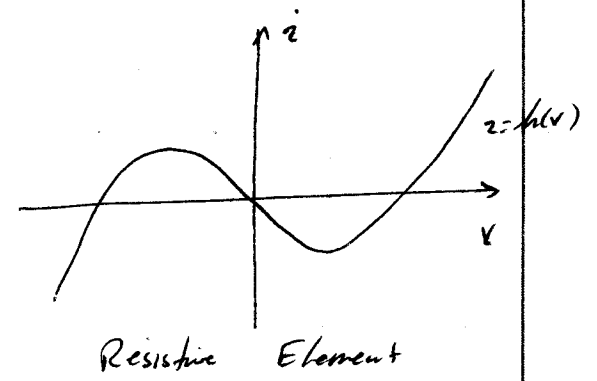
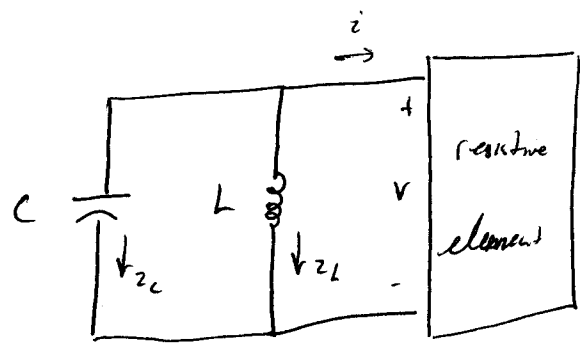
$$\dot{x}_2 = -\frac{k}{m} x_1 - \frac{c}{m} x_2 + M_0 g$$

which are both linear (or affine).

Therefore the system is piecewise linear.

Example: Negative Resistance Oscillator

Consider the circuit



We can write

$$z_c + z_L + z = 0$$

or

$$C \frac{dv}{dt} + \frac{1}{L} \int_{-\infty}^t v(s) ds + h(v) = 0$$

or

$$LC \frac{d^2v}{dt^2} + v + L h'(v) \frac{dv}{dt} = 0 \quad \text{where } h'(v) = \frac{\partial h}{\partial v}(v)$$

Changing the time scale to $\tau = \frac{t}{\sqrt{LC}} \Rightarrow d\tau = \frac{dt}{\sqrt{LC}}$
 $\Rightarrow (d\tau)^2 = d\tau(d\tau) = \frac{dt^2}{LC}$

~~and~~ and defining $\dot{v} = \frac{dv}{d\tau}$ gives

$$\ddot{v} + \epsilon h'(v) \dot{v} + v = 0 \quad \text{where } \epsilon = \sqrt{LC}$$

Setting $x_1 = v$ and $x_2 = \dot{v}$ we get

$$\dot{x}_1 = x_2$$

$$\dot{x}_2 = -x_1 - \epsilon h'(x_1) x_2$$

We could have also chosen the state variables

$$\text{as } z_1 = z_L \quad z_2 = v_L$$

then

$$\frac{dz_1}{dt} = \frac{dz_L}{dt} = \frac{1}{L} \frac{dv_L}{dt} = \frac{1}{L} z_2$$

$$\frac{dz_2}{dt} = \frac{dv_L}{dt} = -\frac{1}{C} z_L - \frac{1}{C} h(v) = -\frac{1}{C} [z_1 + h(z_2)]$$

again letting $\tau = \frac{t}{\sqrt{LC}}$ gives

$$\dot{z}_1 = \frac{1}{\varepsilon} z_2$$

$$\dot{z}_2 = -\varepsilon [z_1 + h(z_2)]$$

While the state variables are different, we are representing the same physical system, therefore we should be able to transform between the two representations in a smooth way

Note that

$$\begin{aligned} z_1 = z_L &= -z - z_c \\ &= -h(v) - C \frac{dv}{dt} \\ &= -h(x_1) - \sqrt{\frac{C}{L}} \frac{dv}{dt} \\ &= -h(x_1) - \frac{1}{\varepsilon} x_2 \end{aligned}$$

$$z_2 = v_L = x_1$$

$$\therefore z = T(x) = \begin{pmatrix} -h(x_1) - \frac{1}{\varepsilon} x_2 \\ x_1 \end{pmatrix}$$

Also

$$x_1 = v = z_L$$

$$x_2 = \frac{dv}{dt} = \sqrt{LC} \frac{dv}{dt} = \sqrt{LC} \left(-\frac{1}{C} z_L - \frac{1}{C} h(v) \right)$$

$$= -\sqrt{\frac{L}{C}} z_1 - \sqrt{\frac{L}{C}} h(z_2)$$

$$\Rightarrow \pi = T^{-1}(z) = \begin{pmatrix} z_2 \\ -\varepsilon z_1 - \varepsilon h(z_2) \end{pmatrix}$$

T is a continuous transformation with a continuous inverse. Such transformations are called diffeomorphisms and play an important role in nonlinear analysis (similar to similarity transformations in linear systems theory).