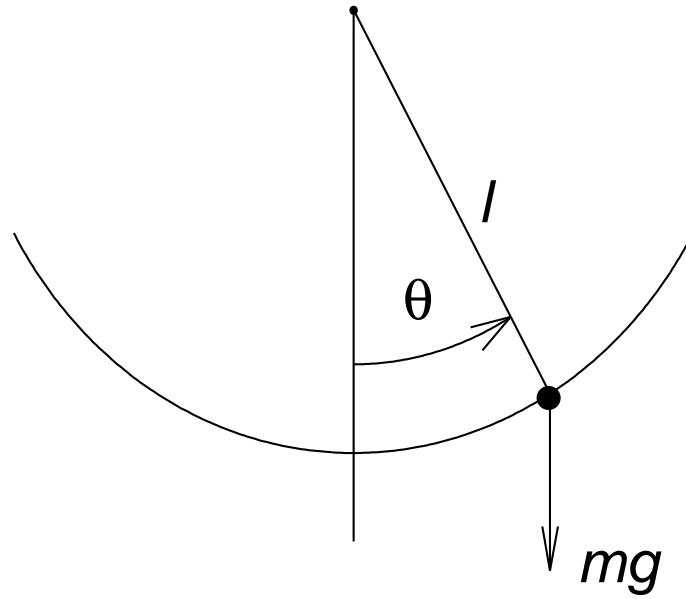


Nonlinear Systems and Control

Lecture # 2

Examples of Nonlinear Systems

Pendulum Equation



$$ml\ddot{\theta} = -mg \sin \theta - kl\dot{\theta}$$

$$x_1 = \theta, \quad x_2 = \dot{\theta}$$

$$\begin{aligned}\dot{x}_1 &= x_2 \\ \dot{x}_2 &= -\frac{g}{l} \sin x_1 - \frac{k}{m} x_2\end{aligned}$$

Equilibrium Points:

$$\begin{aligned}0 &= x_2 \\ 0 &= -\frac{g}{l} \sin x_1 - \frac{k}{m} x_2\end{aligned}$$

$(n\pi, 0)$ for $n = 0, \pm 1, \pm 2, \dots$

Nontrivial equilibrium points at $(0, 0)$ and $(\pi, 0)$

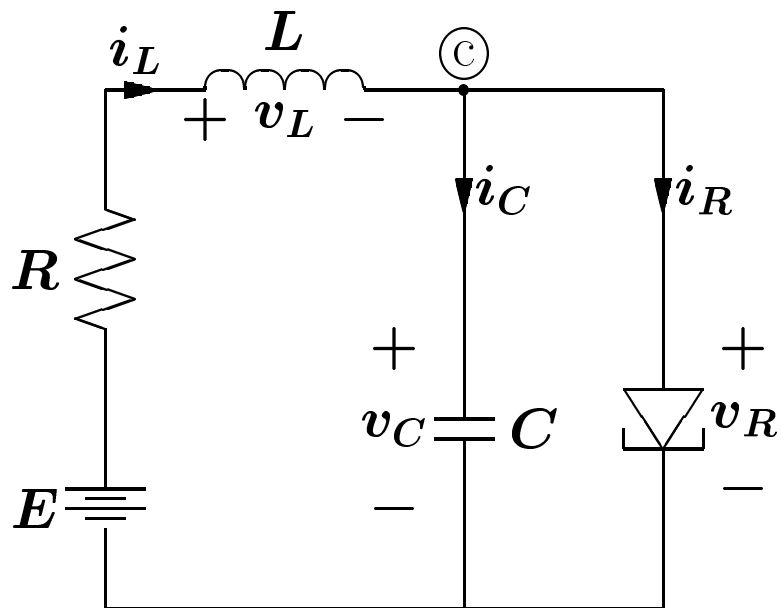
Pendulum without friction:

$$\begin{aligned}\dot{x}_1 &= x_2 \\ \dot{x}_2 &= -\frac{g}{l} \sin x_1\end{aligned}$$

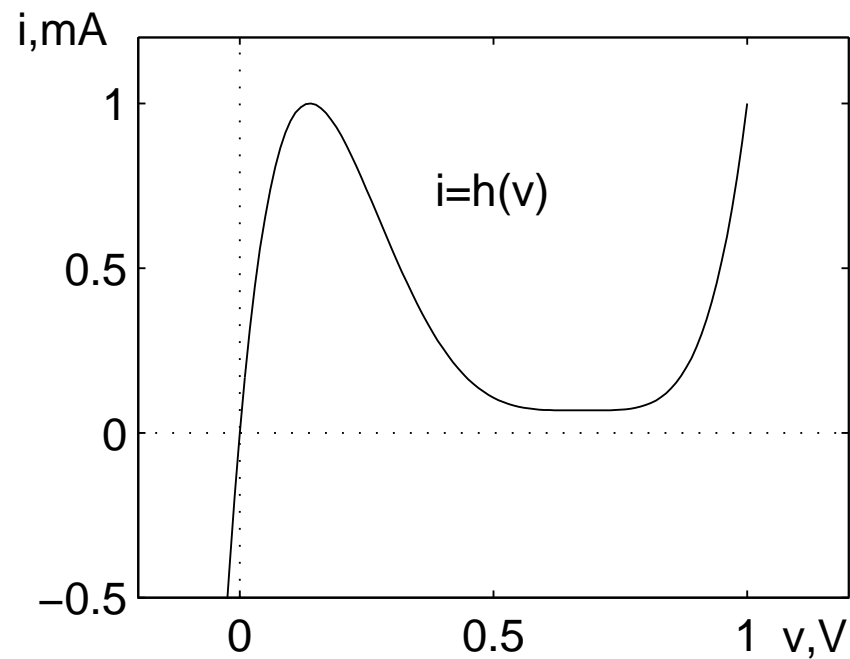
Pendulum with torque input:

$$\begin{aligned}\dot{x}_1 &= x_2 \\ \dot{x}_2 &= -\frac{g}{l} \sin x_1 - \frac{k}{m} x_2 + \frac{1}{ml^2} T\end{aligned}$$

Tunnel-Diode Circuit



(a)



(b)

$$i_C = C \frac{dv_C}{dt}, \quad v_L = L \frac{di_L}{dt}$$

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

$$i_C + i_R - i_L = 0 \Rightarrow i_C = -h(x_1) + x_2$$

$$v_C - E + Ri_L + v_L = 0 \Rightarrow v_L = -x_1 - Rx_2 + u$$

$$\dot{x}_1 = \frac{1}{C} [-h(x_1) + x_2]$$

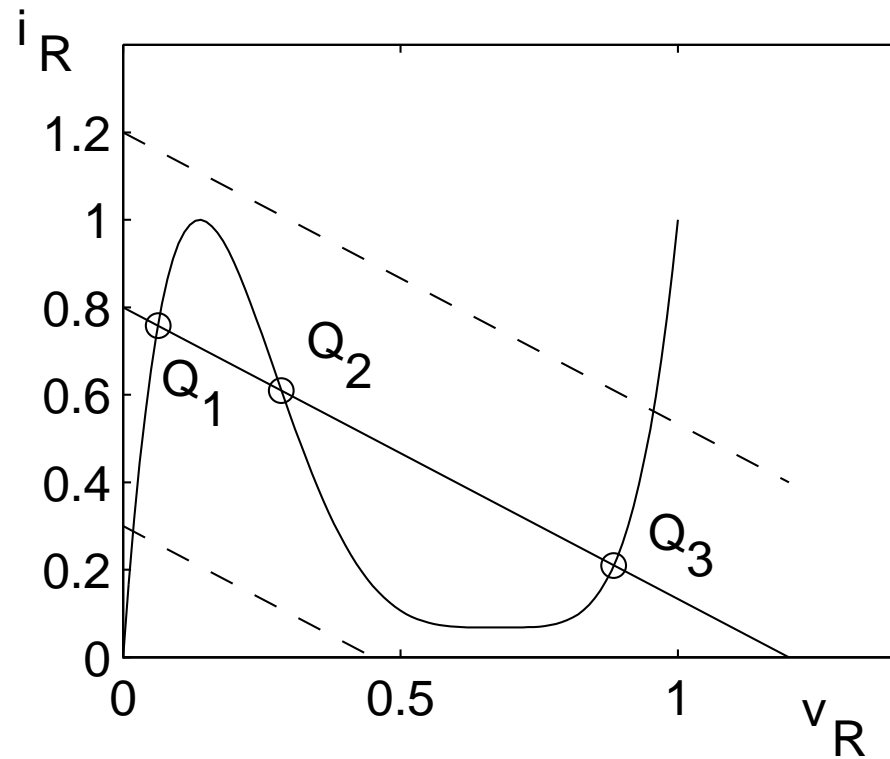
$$\dot{x}_2 = \frac{1}{L} [-x_1 - Rx_2 + u]$$

Equilibrium Points:

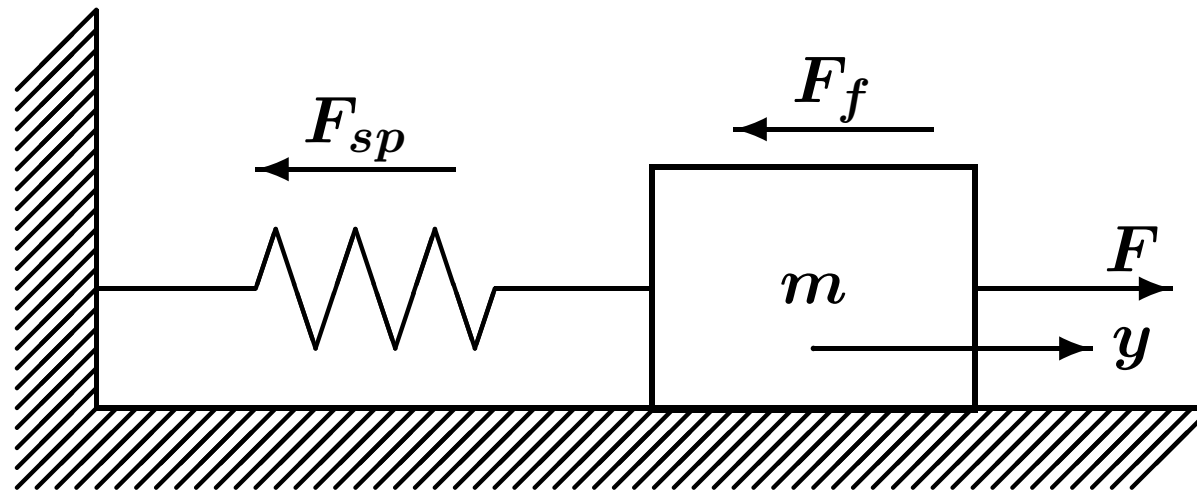
$$0 = -h(x_1) + x_2$$

$$0 = -x_1 - Rx_2 + u$$

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



Mass–Spring System



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

Sources of nonlinearity:

- Nonlinear spring restoring force $F_{sp} = g(y)$
- Static or Coulomb friction

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

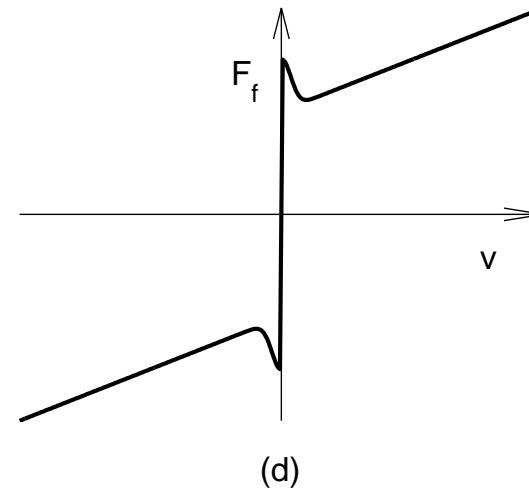
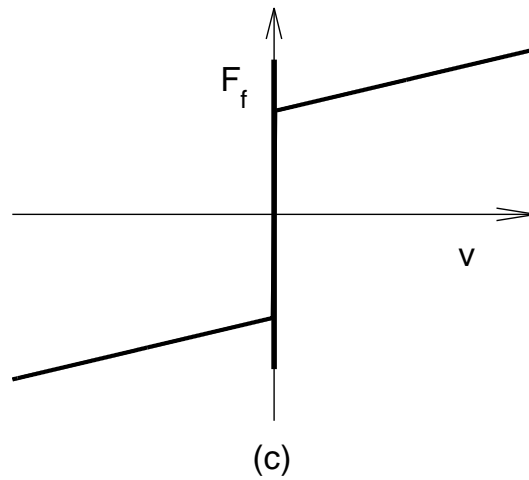
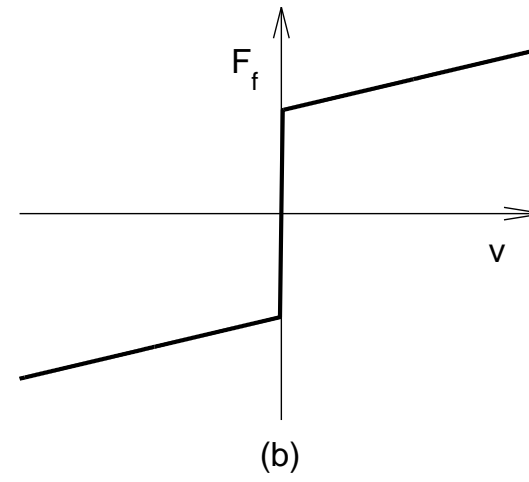
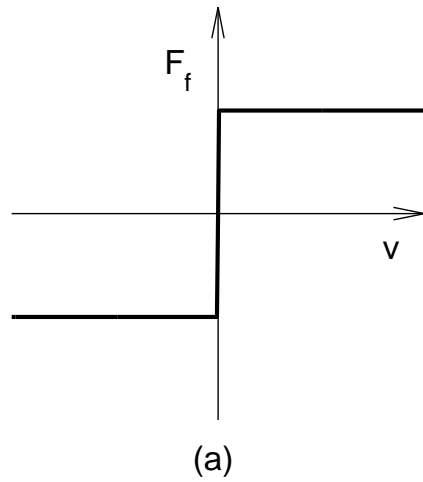
$$g(y) = k(1 - a^2 y^2)y, \quad |ay| < 1 \quad (\text{softening spring})$$

$$g(y) = k(1 + a^2 y^2)y \quad (\text{hardening spring})$$

F_f may have components due to static, Coulomb, and viscous friction

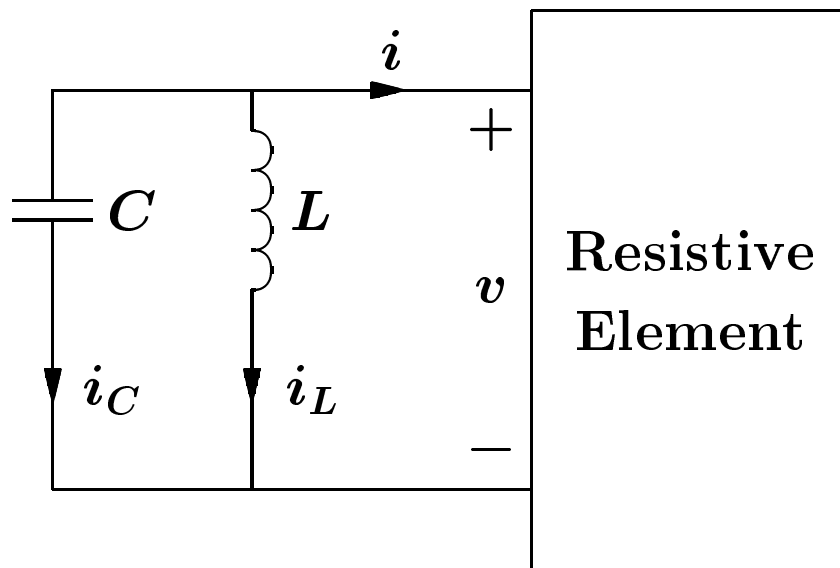
When the mass is at rest, there is a static friction force F_s that acts parallel to the surface and is limited to $\pm \mu_s mg$ ($0 < \mu_s < 1$). F_s takes whatever value, between its limits, to keep the mass at rest

Once motion has started, the resistive force F_f is modeled as a function of the sliding velocity $v = \dot{y}$

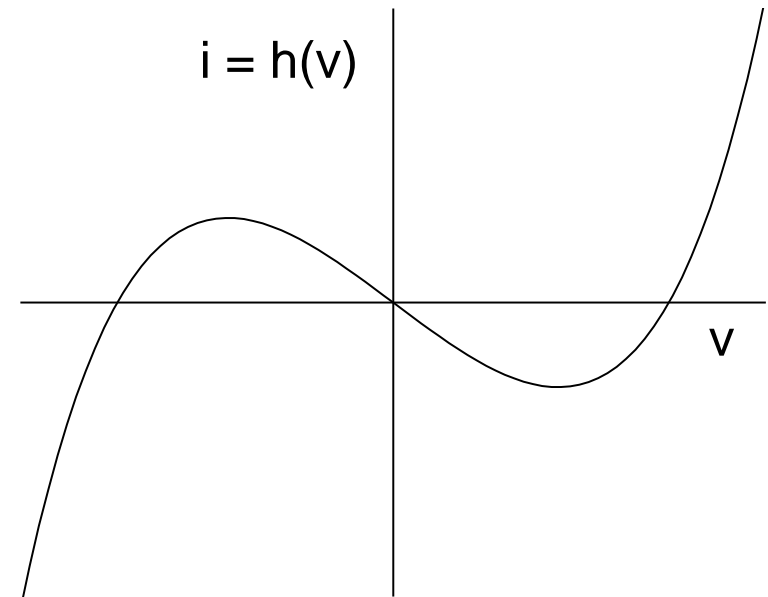


(a) Coulomb friction; (b) Coulomb plus linear viscous friction; (c) static, Coulomb, and linear viscous friction; (d) static, Coulomb, and linear viscous friction—Stribeck effect

Negative-Resistance Oscillator



(a)



(b)

$$h(0) = 0, \quad h'(0) < 0$$

$$h(v) \rightarrow \infty \text{ as } v \rightarrow \infty, \text{ and } h(v) \rightarrow -\infty \text{ as } v \rightarrow -\infty$$

$$i_C + i_L + i = 0$$

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

Differentiating with respect to t and multiplying by L :

$$CL \frac{d^2v}{dt^2} + v + Lh'(v) \frac{dv}{dt} = 0$$

$$\tau = t / \sqrt{CL}$$

$$\frac{dv}{d\tau} = \sqrt{CL} \frac{dv}{dt}, \quad \frac{d^2v}{d\tau^2} = CL \frac{d^2v}{dt^2}$$

Denote the derivative of v with respect to τ by \dot{v}

$$\ddot{v} + \varepsilon h'(v)\dot{v} + v = 0, \quad \varepsilon = \sqrt{L/C}$$

Special case: Van der Pol equation

$$h(v) = -v + \frac{1}{3}v^3$$

$$\ddot{v} - \varepsilon(1 - v^2)\dot{v} + v = 0$$

State model: $x_1 = v, \quad x_2 = \dot{v}$

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

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

Another State Model: $z_1 = i_L, \quad z_2 = v_C$

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

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

Change of variables: $z = T(x)$

$$x_1 = v = z_2$$

$$\begin{aligned} x_2 &= \frac{dv}{d\tau} = \sqrt{CL} \frac{dv}{dt} = \sqrt{\frac{L}{C}} [-i_L - h(v_C)] \\ &= \varepsilon [-z_1 - h(z_2)] \end{aligned}$$

$$T(x) = \begin{bmatrix} -h(x_1) - \frac{1}{\varepsilon} x_2 \\ x_1 \end{bmatrix}, \quad T^{-1}(z) = \begin{bmatrix} z_2 \\ -\varepsilon z_1 - \varepsilon h(z_2) \end{bmatrix}$$

Adaptive Control

Plant : $\dot{y}_p = a_p y_p + k_p u$

Reference Model : $\dot{y}_m = a_m y_m + k_m r$

$$u(t) = \theta_1^* r(t) + \theta_2^* y_p(t)$$

$$\theta_1^* = \frac{k_m}{k_p} \quad \text{and} \quad \theta_2^* = \frac{a_m - a_p}{k_p}$$

When a_p and k_p are unknown, we may use

$$u(t) = \theta_1(t)r(t) + \theta_2(t)y_p(t)$$

where $\theta_1(t)$ and $\theta_2(t)$ are adjusted on-line

Adaptive Law (gradient algorithm):

$$\dot{\theta}_1 = -\gamma(y_p - y_m)r$$

$$\dot{\theta}_2 = -\gamma(y_p - y_m)y_p, \quad \gamma > 0$$

State Variables: $e_o = y_p - y_m$, $\phi_1 = \theta_1 - \theta_1^*$, $\phi_2 = \theta_2 - \theta_2^*$

$$\dot{y}_m = a_p y_m + k_p(\theta_1^* r + \theta_2^* y_m)$$

$$\dot{y}_p = a_p y_p + k_p(\theta_1 r + \theta_2 y_p)$$

$$\dot{e}_o = a_p e_o + k_p(\theta_1 - \theta_1^*)r + k_p(\theta_2 y_p - \theta_2^* y_m)$$

$$= \dots\dots\dots + k_p[\theta_2^* y_p - \theta_2^* y_p]$$

$$= (a_p + k_p \theta_2^*)e_o + k_p(\theta_1 - \theta_1^*)r + k_p(\theta_2 - \theta_2^*)y_p$$

Closed-Loop System:

$$\dot{e}_o = a_m e_o + k_p \phi_1 r(t) + k_p \phi_2 [e_o + y_m(t)]$$

$$\dot{\phi}_1 = -\gamma e_o r(t)$$

$$\dot{\phi}_2 = -\gamma e_o [e_o + y_m(t)]$$