

System Modeling

Definitions

	relevant	not relevant
System Modeling (grey/white Box)	VS.	System Identification (black Box)
Parametric Model (ODE, PDE, transfer func., ...)	VS.	Non-Parametric Model (lists, tables, ...)
Forward Formulation (cause \rightarrow effect)	VS.	Backward Formulation (effect \rightarrow cause)
Relevant Dynamics (need model)	VS.	Fast Dynamics (algebraic) VS. Slow Dynamics (constant)

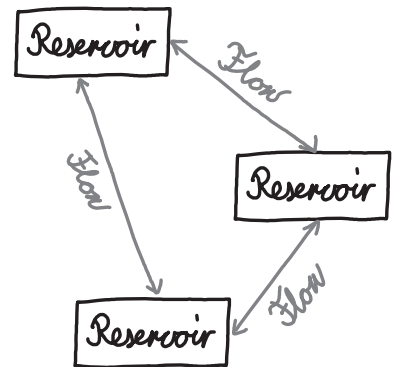
Reservoir-Based Approach

- Reservoir: thermal/kinetic energy, mass, information, ... (tracked by state variable)
- Flows: heat, mass, etc. flowing between reservoirs (typically driven by reservoir difference)

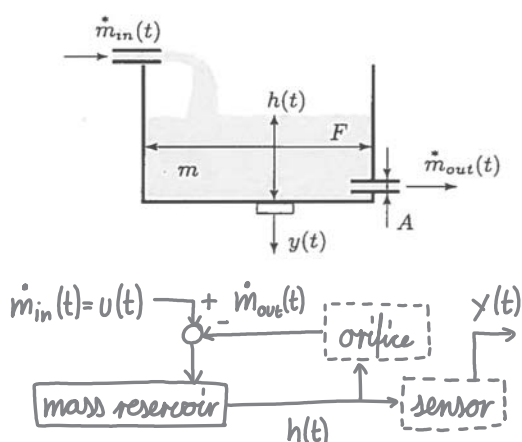
GUIDELINES:

- 1 define system boundaries (inputs, outputs)
- 2 identify relevant reservoirs (state variable)
- 3 formulate conservation law:

$$\frac{d}{dt}(\text{reservoir content}) = \sum(\text{inflows}) - \sum(\text{outflows})$$
- 4 formulate algebraic relations expressing flows between reservoirs
- 5 simplify as much as possible
- 6 identify unknown system parameters with experiments
- 7 use other experiments to validate model



Reservoir-Based Approach Example



- 1 system input: $\dot{m}_{in}(t)$; system output: $h(t)$
- 2 system reservoir: $m(t)$
- 3 mass balance: $\frac{d}{dt} m(t) = \dot{m}_{in}(t) - \dot{m}_{out}(t)$
- 4 relations: $\dot{m}_{out}(t) = A_s \sqrt{2gh(t)}$ (Bernoulli)
 $m(t) = \rho F h(t)$
- 5 simplify: $\rho F \frac{d}{dt} h(t) = \dot{m}_{in}(t) - A_s \sqrt{2gh(t)}$

Basic Modeling Elements

• Mechanical Systems (2D, reservoir: energy, 1DOF)

$$\begin{aligned}
 T_t(t) &= \frac{1}{2} m (\dot{x}_{c.g.}(t) + \dot{y}_{c.g.}(t))^2 \quad (\text{translational energy}) \\
 T_r(t) &= \frac{1}{2} \Theta \omega^2(t) \quad (\text{rotational energy}) \\
 U(t) &= U(x(t), y(t)) = mgx \quad (\text{potential energy})
 \end{aligned}
 \left. \begin{array}{l} \\ \\ \end{array} \right\} \begin{array}{l} E(t) = T_t(t) + T_r(t) + U(t) \quad (\text{total energy}) \\ \frac{d}{dt} E(t) = \sum P_i(t) \quad (\text{conservation law}) \\ \text{mech. powers acting on body} \end{array}$$

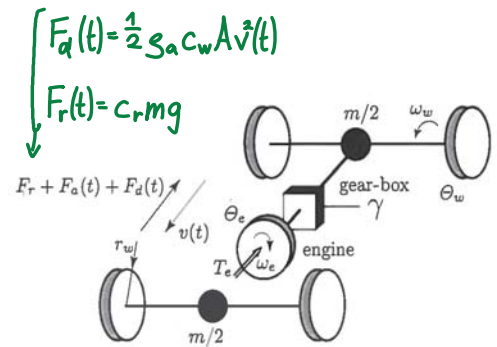
Example 1:

input: $T_e(t)$, output: $y(t) \sim v(t)$, state: $v(t)$

$$\begin{aligned}
 E_{tot} &= \frac{1}{2} m v^2(t) + 4 \frac{1}{2} \Theta_w \omega_w^2(t) + \frac{1}{2} \Theta_e \omega_e^2(t) \\
 &= \frac{1}{2} \left(m + \frac{4 \Theta_w}{r_w^2} + \frac{\Theta_e}{\gamma^2 r_w^2} \right) v^2(t) \quad \left(\begin{array}{l} v(t) = r_w \omega_w(t) \\ = r_w \gamma \omega_e(t) \end{array} \right)
 \end{aligned}$$

$$\frac{d}{dt} E_{tot}(t) = T_e(t) \omega_e(t) - F_r v(t) - F_a(t) v(t) - F_d(t) v(t)$$

$$\hookrightarrow \frac{d}{dt} v(t) = \text{func}(T_e(t), v(t))$$

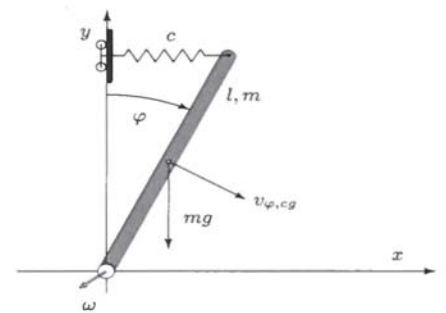


Example 2:

input: none, output: $y(t) \sim \varphi(t)$, state: $\varphi(t), \dot{\varphi}(t)$

$$E_{tot}(t) = \underbrace{\frac{1}{2} m \left(\frac{L \dot{\varphi}(t)}{2} \right)^2}_{T_{trans}} + \underbrace{\frac{1}{2} \frac{m L^2}{12} \dot{\varphi}^2(t)}_{T_{rot}} + \underbrace{mg \frac{L}{2} \cos(\varphi(t))}_{U_{grav}} + \underbrace{\frac{1}{2} c (L \sin(\varphi(t)))^2}_{U_{spring}}$$

$$\frac{d}{dt} E_{tot}(t) = 0 \rightarrow \frac{1}{3} m L^2 \ddot{\varphi}(t) = \left(\frac{L}{2} mg - c L^2 \cos(\varphi(t)) \right) \sin(\varphi(t))$$



Example 3:

input: T_1, T_2 output: ω_1

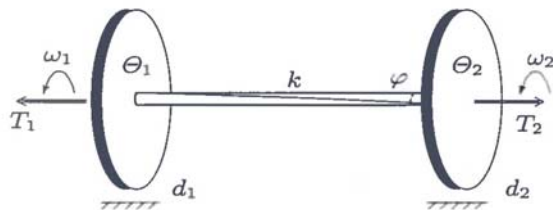
state: $\omega_1, \omega_2, \varphi$

reservoir: compressor, turbine, shaft (mech. energy)

$$\left\{ \begin{array}{l} \frac{d}{dt} \left(\frac{1}{2} \Theta_1 \omega_1^2(t) \right) = -P_1(t) - P_2(t) + P_3(t) \\ \frac{d}{dt} \left(\frac{1}{2} \Theta_2 \omega_2^2(t) \right) = -P_4(t) - P_5(t) + P_6(t) \\ \frac{d}{dt} \left(\frac{1}{2} k \varphi^2(t) \right) = -P_3(t) + P_4(t) \end{array} \right. \quad \begin{array}{l} (\text{compressor balance}) \\ (\text{turbine balance}) \\ (\text{shaft balance}) \end{array}$$

$$\left\{ \begin{array}{l} \Theta_1 \frac{d}{dt} \omega_1(t) = -T_1(t) - d_1 \omega_1(t) + k \varphi(t) \\ \Theta_2 \frac{d}{dt} \omega_2(t) = T_2(t) - d_2 \omega_2(t) - k \varphi(t) \\ \frac{d}{dt} \varphi(t) = \omega_2(t) - \omega_1(t) \end{array} \right.$$

$$\begin{array}{ll} P_1 = T_1 \omega_1 \quad (\text{input}) & P_4 = \omega_2 k \varphi \quad (\text{spring}) \\ P_2 = d_1 \omega_1^2 \quad (\text{fric. loss}) & P_5 = d_2 \omega_2^2 \quad (\text{fric. loss}) \\ P_3 = \omega_1 k \varphi \quad (\text{spring}) & P_6 = T_2 \omega_2 \quad (\text{input}) \end{array}$$



• Mechanical systems (2D, reservoir: energy, DOF=n) (Lagrange Method)

- 1) Define inputs and outputs $u(t), y(t)$
- 2) Define generalized coordinates $q(t) = (q_1(t), q_2(t), \dots, q_n(t))$, $\dot{q} = (\dots)$
- 3) Build the Lagrange function $L(q, \dot{q}) = \sum_{i=1}^n T_i(q, \dot{q}) - U_i(q)$
- 4) Build n system dynamics equations $\frac{d}{dt} \left\{ \frac{\partial L}{\partial \dot{q}_k} \right\} - \frac{\partial L}{\partial q_k} = Q_k \quad (k=1, \dots, n)$
- 4+ If system has constraints: replace 4 $\frac{d}{dt} \left\{ \frac{\partial L}{\partial \dot{q}_k} \right\} - \frac{\partial L}{\partial q_k} - \sum_{j=1}^v \mu_j \alpha_{j,k} = Q_k$
 $\sum_{k=1}^n \alpha_{j,k} \cdot \dot{q}_k(t) = 0 \quad \leftarrow (j=1, \dots, v) \uparrow (k=1, \dots, n)$
- 5) Solve equation system for $\ddot{q}_k \quad (k=1, \dots, n)$ and $\mu_j \quad (j=1, \dots, v)$
 \rightarrow or inertia Matrix form: $M(q) \ddot{q} = f(q, \dot{q}, Q)$

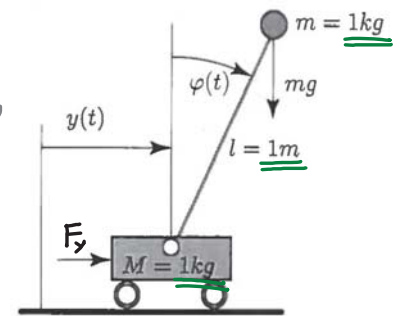
n : # DOF
 m : # Bodies
 q_k : angle or distance coordinate
 Q_k : force or torque on q_k
 v : # constraints
 μ : Lagrange multiplier
 α : constants

Example 1:

$m=2, n=2$; $q_1=\gamma, q_2=\varphi$; input: F_y output: ?

$$L(q, \dot{q}) = \underbrace{\frac{1}{2} \dot{\gamma}^2(t)}_{T_1(t)} - \underbrace{0}_{U_1(t)} + \underbrace{\frac{1}{2} \dot{\gamma}^2(t) + \cos(\varphi(t)) \dot{\varphi}(t) \dot{\gamma}(t) + \frac{1}{2} \dot{\varphi}^2(t)}_{T_2(t)} - \underbrace{\cos(\varphi(t)) g}_{U_2(t)}$$

$$\begin{cases} \rightarrow 2\ddot{\gamma}(t) + \ddot{\varphi}(t) \cos(\varphi(t)) - \dot{\varphi}^2(t) \sin(\varphi(t)) = F_y(t) \\ \rightarrow \ddot{\varphi}(t) + \ddot{\gamma}(t) \cos(\varphi(t)) - g \sin(\varphi(t)) = 0 \end{cases} \quad \begin{matrix} \rightarrow \ddot{\gamma}(t) = \dots \\ \rightarrow \ddot{\varphi}(t) = \dots \end{matrix}$$



Example 2:

$n=3, m=2, v=1$; $q_1=\psi, q_2=\chi, q_3=\varphi$; input: $u=T$, output: $y=(R+r)\sin(\chi)$

$$L(q, \dot{q}) = T_{tot}(q, \dot{q}) - U_{tot}(q, \dot{q}) = \frac{1}{2} m(R+r)^2 \dot{\chi}^2(t) + \frac{1}{2} v \dot{\varphi}^2(t) + \frac{1}{2} \Theta \dot{\psi}^2(t) - (-mg(1 - \cos(\chi(t))) \cdot (R+r))$$

$$Q_1 = u(t), Q_2 = Q_3 = 0 \quad (u(t): \text{input torque})$$

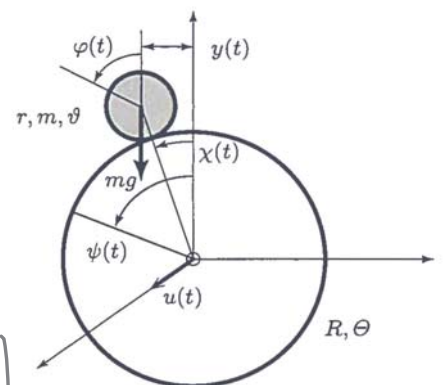
$$R \dot{\psi}(t) - (R+r) \dot{\chi}(t) + r \dot{\varphi}(t) = 0 \quad (\text{no-slip condition})$$

$$\alpha_1 \dot{q}_1 + \alpha_2 \dot{q}_2 + \alpha_3 \dot{q}_3 = 0 \rightarrow \alpha_1 = R, \alpha_2 = -R-r, \alpha_3 = r$$

\rightarrow system equations: (for $\ddot{\psi}, \ddot{\chi}, \ddot{\varphi}, \mu$)

\rightarrow simplify: (eliminate $\ddot{\varphi}, \mu$)

$$\begin{pmatrix} \Theta + v \frac{R^2}{r^2} & -v \frac{R(R+r)}{r^2} \\ -v \frac{R(R+r)}{r^2} & m(R+r)^2 + v \frac{(R+r)^2}{r^2} \end{pmatrix} \begin{pmatrix} \ddot{\psi} \\ \ddot{\chi} \end{pmatrix} = \begin{pmatrix} u \\ mg(R+r) \sin(\chi) \end{pmatrix}$$



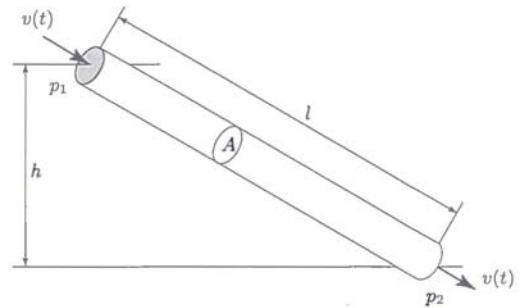
• Hydraulic Systems

duct element: (momentum conservation)

$$\underbrace{\rho A}_{m} \frac{d}{dt} v(t) = A(p_1(t) - p_2(t)) + A \rho g h - F_f(t)$$

$$F_f(t) = A \cdot \lambda(v(t)) \cdot \frac{L}{d} \cdot \frac{\rho}{2} \cdot \text{sign}(v(t)) \cdot v^2(t)$$

λ : from Moody-Diagram



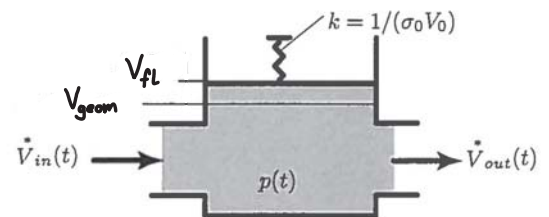
compress. tank: (mass conservation)

$$\frac{d}{dt} m(t) = \dot{m}_{in}(t) - \dot{m}_{out}(t)$$

$$\frac{1}{\rho} \frac{d}{dt} \rho V_{fl}(t) = \dot{V}_{in}(t) - \dot{V}_{out}(t) \quad (\dot{V} = \frac{\dot{m}}{\rho}, V_{fl} \neq V_{geom})$$

$$\hookrightarrow p(t) = \frac{1}{\sigma_0} \frac{V_{fl}(t) - V_{geom}}{V_{geom}} + p_{stat, geom}$$

(σ_0 : compressibility)

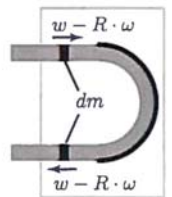
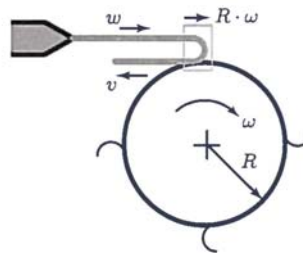


pelton turbine: (momentum conservation)

$$d(mv) = 2(w - R\omega) dm = 2(w - R\omega) \dot{V}_s dt$$

$$\hookrightarrow T_T = \frac{d(mv)}{dt} R = 2 \rho R (w - R\omega) \dot{V}$$

$$\hookrightarrow (T_T)_{max} \text{ for } R = \frac{1}{2} w / \omega$$



• Fluiddynamic systems

• Valves:

$$\dot{m}(t) = c_d A \sqrt{2g(p_{in}(t) - p_{out}(t))} \quad (\text{incompressible / Bernoulli})$$

$$\dot{m}(t) = c_d A \frac{p_{in}(t)}{\sqrt{RT_{in}(t)}} \cdot \psi(\Pi = p_{out}(t)/p_{in}(t)) \quad (\text{compressible / isenthalpic} \rightarrow T_{in} = T_{out})$$

$$\begin{array}{l} \text{real} \left\{ \begin{array}{ll} \sqrt{\left(\frac{2K}{K+1}\right)^{\frac{K+1}{K-1}}} & \text{for } \Pi < \Pi_{cr} \\ \Pi^{-1/K} \cdot \sqrt{\frac{2K}{K-1} \cdot \left(1 - \Pi^{\frac{K-1}{K}}\right)} & \text{for } \Pi \geq \Pi_{cr} \end{array} \right. \\ \Pi_{cr} = \left(\frac{2}{K+1}\right)^{\frac{K}{K-1}} \end{array} \quad \begin{array}{l} \text{approx.} \left\{ \begin{array}{ll} 1/\sqrt{2} & \text{for } \Pi < 1/2 \\ \sqrt{2\Pi \cdot (1-\Pi)} & \text{for } \Pi_{tr} > \Pi \geq 1/2 \\ a \cdot (\Pi-1)^3 + b(\Pi-1) & \text{for } \Pi > \Pi_{tr} \end{array} \right. \\ a = \frac{\psi'(\Pi_{tr}) \cdot (\Pi_{tr}-1) - \psi(\Pi_{tr})}{2(\Pi_{tr}-1)^3}; \quad b = \psi'(\Pi) - 3a(\Pi_{tr}-1)^2 \end{array}$$

• Gas Turbines:

$$p_3 = p_{in}, p_4 = p_{out}, T_3 = T_{in}, \omega_t \rightarrow T_4 = T_{out}, \dot{m}_t, T_{t,torg}$$

$$T_4 = T_3 \left(1 - \eta_t \left(1 - \Pi_t^{\frac{1-K}{K}}\right)\right)$$

$$\Pi_t = p_3/p_4$$

$$\dot{m}_t = \dot{m}_t \frac{p_3/p_{3ref}}{\sqrt{T_3/T_{3ref}}}$$

$$\eta_t = \text{table}(c_u) \leftarrow c_u = \frac{r_t \omega_t}{c_{us}} \leftarrow c_{us} = \sqrt{2c_p T_3 \left(1 - \Pi_t^{\frac{1-K}{K}}\right)}$$

$$T_{t,torg} = \frac{\eta_t \dot{m}_t c_p T_3}{\omega_t} \left(1 - \Pi_t^{\frac{1-K}{K}}\right)$$

$$\dot{m}_t = \text{table}(\Pi_t, \text{nozzle_pos}, \omega_t)$$

• Gas Compressor:

$$p_1 = p_{in}, p_2 = p_{out}, T_1 = T_{in}, \omega_c \rightarrow T_2 = T_{out}, \dot{m}_c, T_{c,torg}$$

$$T_2 = T_1 \left(1 - \frac{1}{\eta_c} \left(\Pi_c^{\frac{1-K}{K}} - 1\right)\right)$$

$$\Pi_c = p_2/p_1$$

$$\dot{m}_c = \dot{m}_c \frac{p_1/p_{1ref}}{\sqrt{T_1/T_{1ref}}}$$

$$\eta_c = \text{table}(\Pi_c, \omega_c)$$

$$T_{c,torg} = \frac{\dot{m}_c c_p T_1}{\eta_c \omega_c} \left(\Pi_c^{\frac{1-K}{K}} - 1\right)$$

$$\dot{m}_c = \text{table}(\Pi_c, \omega_c)$$

• Thermodynamic systems (reservoir: internal energy U)

$$d/dt U(t) = \dot{Q} - \dot{W} + \dot{H}_{in} - \dot{H}_{out}$$

$$\left\{ \begin{array}{l} U(T) = m \cdot \int_0^T c_v(T) dT = c_v \cdot m \cdot T \quad (\text{for liquid/solid, } c = c_p = c_v) \\ \dot{H}(t) = \dot{Q}(t)_{isobar} = c_p \cdot \dot{m} \cdot T \end{array} \right. \quad (\text{for ideal gas, } R = c_p - c_v, \quad p = \rho R T)$$

$$pV = nRT$$

• conduction: $\dot{Q} = \kappa A / l (T_1 - T_2)$

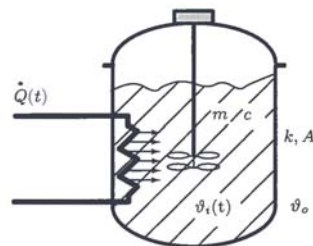
• convection: $\dot{Q} = k A (T_1 - T_2)$

• radiation: $\dot{Q} = \epsilon \sigma A (T_1^4 - T_2^4)$

Example 1:

input: $\dot{Q}_{in}(t)$ reservoir: $U(t)$ level: $T(t) = T_i(t) - T_o$

$$d/dt U(t) = m c d/dt T(t) = \dot{Q}_{in}(t) - \underbrace{k A T(t)}_{\dot{Q}_{out}(t)}$$



Example 3:

inputs: $\dot{m}_{in/out}, \dot{H}_{in/out}$, reservoirs: $m(t), U(t)$

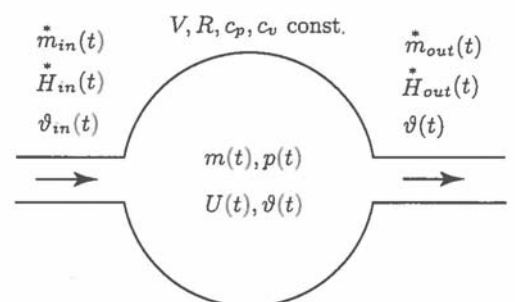
$$d/dt U(t) = \dot{H}_{in}(t) - \dot{H}_{out}(t)$$

$$d/dt m(t) = \dot{m}_{in}(t) - \dot{m}_{out}(t)$$

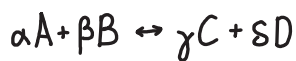
using $U = c_v m T, \dot{H} = c_p \dot{m} T, m = pV/RT$ and combining prev. eq.:

$$d/dt T = \frac{TR}{pV c_v} (c_p \dot{m}_{in} T_{in} - c_p \dot{m}_{out} T - (\dot{m}_{in} - \dot{m}_{out}) c_v T)$$

$$d/dt p = \frac{c_p R}{c_v V} (\dot{m}_{in} T_{in} - \dot{m}_{out} T)$$



• Chemical systems



$$([\cdot] = \frac{\text{mol}}{\text{m}^3})$$

$$(k: \text{factor } E: \text{activation } E)$$

$$\begin{aligned} \frac{d}{dt}[A] &= \alpha \cdot (r^- [C]^\gamma [D]^\delta - r^+ [A]^\alpha [B]^\beta) + \dot{m}_A \cdot \frac{1}{V M_A} + \dots \\ \frac{d}{dt}[C] &= \gamma (r^+ [A]^\alpha [B]^\beta - r^- [C]^\gamma [D]^\delta) + \dot{m}_C \cdot \frac{1}{V M_C} + \dots \end{aligned} \quad \begin{cases} r^+ = k^+(T, p, \dots) \cdot e^{-E^+/RT} & \text{if } \rightarrow \text{ then } r^+ = 0 \\ r^- = k^-(T, p, \dots) \cdot e^{-E^-/RT} & \text{if } \leftarrow \text{ then } r^- = 0 \end{cases}$$

$$\frac{d}{dt}U = Q - \dot{W} + H_A V_{dt}^A [A] + H_B V_{dt}^B [B] + H_C V_{dt}^C [C] + H_D V_{dt}^D [D] + \underbrace{c_p T_{in}}_H \dot{m}_{in} - \underbrace{c_p T_{out}}_H \dot{m}_{out}$$

Example 1:

(assumptions: reaction $A+B \rightarrow C$, $[B]=\text{const.}$, no dissociation $C \rightarrow A+B$, $g \& m \text{ const.}$)

(input: \dot{Q} , output: $[C], T$, disturb: $[A_i], T_i$, reservoirs: n_A, n_C, U)

$$\frac{d}{dt}n_A(t) = \dot{V} \cdot [A_i](t) - \dot{V} [A](t) - V k^- e^{-E/(RT)} \cdot [A](t) \cdot [B]$$

$$\frac{d}{dt}n_C(t) = -\dot{V} [C](t) + V k^- e^{-E/(RT)} [A](t) \cdot [B]$$

$$\frac{d}{dt}U(T, n_A, n_B, n_C) = \dot{H}_i(T_i(t)) - \dot{H}(T(t)) + \dot{Q}(t) \quad (\dot{H}_i(T_i) = \dot{m} \cdot c_p \cdot T_i; \dot{H}(t) = \dot{m} \cdot c_p \cdot T)$$

$$dU(T, n_A, n_B, n_C) = \frac{\partial U}{\partial T} dT + \frac{\partial U}{\partial n_A} dn_A + \frac{\partial U}{\partial n_B} dn_B + \frac{\partial U}{\partial n_C} dn_C$$

$$= \sum V c_v dT + H_A dn_A + H_B dn_B + H_C dn_C \quad (H_{A,B,C}: \text{enthalpy of formation})$$

$$\tau \frac{d}{dt}[A](t) = [A_i](t) - (1 + \tau k^- e^{-E/(RT(t))}) \cdot [A](t)$$

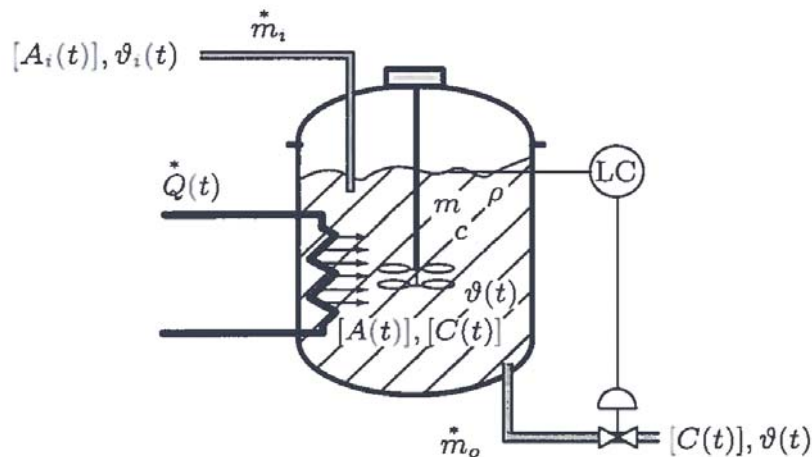
$$\tau \frac{d}{dt}[C](t) = -[C](t) + \tau k^- e^{-E/(RT(t))} \cdot [A](t)$$

$$\tau \frac{d}{dt}T(t) = T_i(t) - T(t) + \frac{1}{\sum c_v} \cdot \frac{\dot{Q}(t)}{\dot{V}} + \tau H_0 \frac{k^-}{\sum c_v} e^{-E/(RT(t))} \cdot [A](t)$$

$$\left(\begin{aligned} \tau &= V/\dot{V} \\ k &= k^- [B] \\ H_0 &= H_A + H_B - H_C \end{aligned} \right)$$

Steady state ($\dot{Q}=0, d/dt[\cdot]=0$):

$$\dot{H}_{flow} + \dot{Q}_{chem} = \dot{m} c_p (T_i - T) + H_0 \frac{V k^- e^{-E/(RT)}}{1 + \tau k^- e^{-E/(RT)}} = 0$$



Distributed Parameter Systems (missing!)

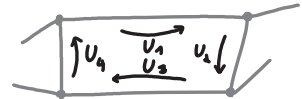
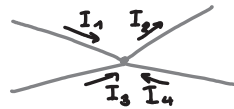
• Electromagnetic Systems (RLC Networks)

	capacitance C	inductance L	resistance R
energy	$W_E = \frac{1}{2} C U(t)^2$	$W_M = \frac{1}{2} L I(t)^2$	$W = 0$
level variable	$U(t)$	$I(t)$	—
conservation law	$C \frac{d}{dt} U(t) = I(t)$	$L \frac{d}{dt} I(t) = U(t)$	—

Ohms law: $U = R \cdot I$

1st Kirchhoff's law: $\sum I_i = 0$

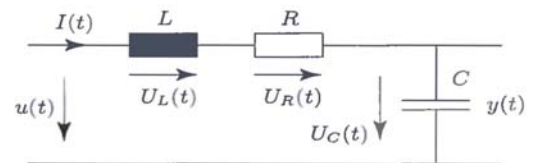
2nd Kirchhoff's law: $\sum U_i = 0$



Example:

input: $u(t)$, output: $U_C(t)$, reservoirs: W_E & W_M

$$\left. \begin{array}{l} U_L(t) = L \cdot \frac{d}{dt} I(t) \\ I(t) = C \cdot \frac{d}{dt} U_C(t) \\ U_R(t) = R \cdot I(t) \end{array} \right\} \rightarrow \left\{ \begin{array}{l} U_L(t) = L \cdot C \cdot \frac{d^2}{dt^2} U_C(t) \\ U_C(t) = U_C(t) \\ U_R(t) = R \cdot C \cdot \frac{d}{dt} U_C(t) \end{array} \right\}$$



$$U_L(t) + U_R(t) + U_C(t) = u(t)$$

$$LC \frac{d^2}{dt^2} U_C(t) + RC \frac{d}{dt} U_C(t) + U_C(t) = u(t)$$

• Electromechanical systems

Lorentz law:

$$\underline{F} = q \underline{E} + q \underline{v} \times \underline{B} = I \underline{L} \times \underline{B}$$



Faraday law:

$$U = - \frac{d}{dt} \underline{B} \cdot \underline{A} = - \underline{v} \cdot (\underline{L} \times \underline{B})$$



Example 1:

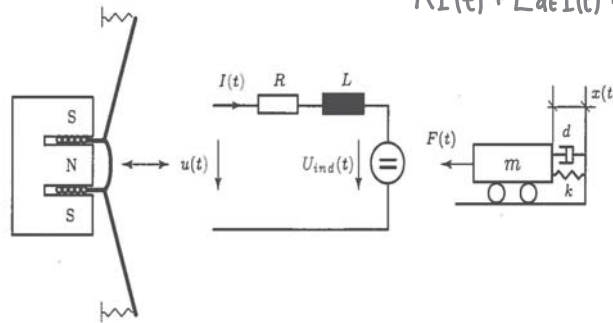
$$F(t) = Bnd\pi \cdot I(t) = \kappa \cdot I(t) \quad (\text{motor law})$$

$$F(t) = m\ddot{x}(t) + d\dot{x}(t) + kx(t) \quad (\text{membrane dynamics})$$

$$U_{ind}(t) = Bnd\pi \cdot \dot{x}(t) = \kappa \cdot \dot{x}(t) \quad (\text{generator law})$$

$$u(t) = U_R(t) + U_L(t) + U_{ind}(t) \quad (\text{RLC dynamics})$$

$$RI(t) + L \frac{d}{dt} I(t) + U_{ind}(t)$$



Example 2:

input: $u(t), T_L(t)$ output: $\omega(t)$ reservoirs: W_{ROT}, W_M

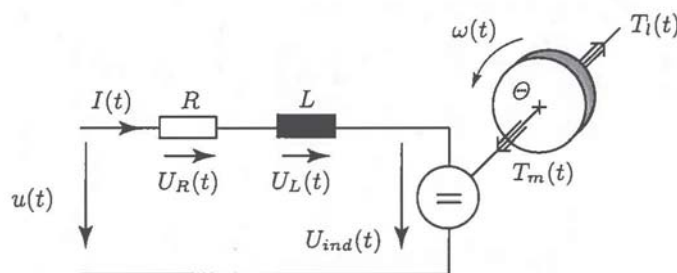
$$L \cdot \frac{d}{dt} I(t) = u(t) - RI(t) - U_{ind}(t) \quad (W_M \text{ conservation})$$

$$\oplus \frac{d}{dt} \omega(t) = T_m(t) - T_L(t) - d \cdot \omega(t) \quad (W_{ROT} \text{ conservation})$$

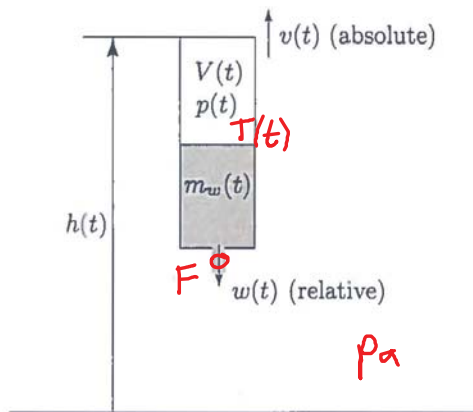
$$T_m(t) = Bnd^2 \frac{\pi}{2} \cdot I(t) = \kappa \cdot I(t) \quad (\text{motor law})$$

$$U_{ind}(t) = Bnd^2 \frac{\pi}{2} \cdot \omega(t) = \kappa \cdot \omega(t) \quad (\text{generator law})$$

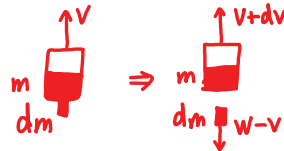
$$\left. \begin{aligned} L \frac{d}{dt} I(t) &= u(t) - RI(t) - \kappa \omega(t) \\ \oplus \frac{d}{dt} \omega(t) &= \kappa I(t) - T_L(t) - d \omega(t) \end{aligned} \right\}$$



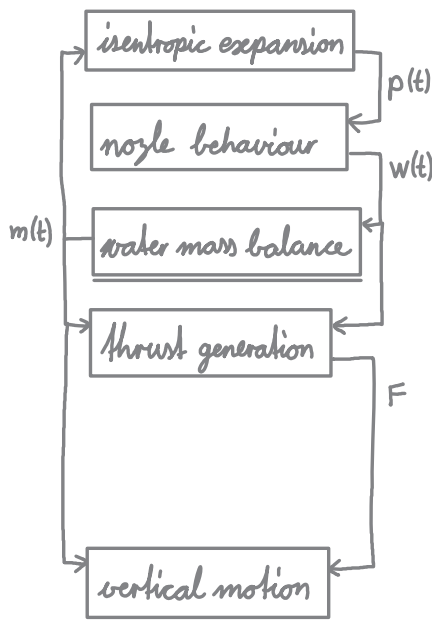
Case Study: Rocket



(assumptions: no drag, isentropic air expansion)
air mass neglected



Water Thrust Phase



$$p(t) = (V(0)/V_L - m_w(t)/s)^K \cdot p(0) \quad (\text{isentropic relation})$$

$$w(t) = \sqrt{2/s (p(t) - p_a)} \quad (\text{Bernoulli eq.})$$

$$d/dt m(t) = s F \cdot w(t) \quad (\text{mass balance})$$

$$dP = m(v+dv) - dm(w-v) - (m+dm)v = m dv - dm w$$

$$dP = -g m dt \quad dm = s F w dt \quad \left. \begin{array}{l} \\ \end{array} \right\} \text{momentum balance}$$

$$\hookrightarrow m(t) d/dt v(t) = F(t) = -g m(t) + s F w^2(t)$$

$$d/dt v(t) = F(t)/m(t) \quad ; \quad d/dt h(t) = v(t)$$

Air Thrust Phase

$$p(t) = m_{\text{air}}(t) R T(t) / V_L \quad \leftarrow \quad d/dt U(t) = c_v m_{\text{air}}(t) d/dt T(t) = \dot{H}_{\text{out}} = -\dot{m}_{\text{out}}(t) c_p T(t)$$

$$w(t) = \text{see valve formula for gas.}$$

$$T(t_1) = p(t_1) \cdot V_L / R m_{\text{air}}(t_1)$$

$$m_{\text{air}}(t_1) = m_{\text{air}}(0) = \frac{p(0) V(0)}{R T(0)}$$

rest is the same as for water phase!

Balistic Phase

$$F(t) = -g m(t) \quad ; \quad m(t) = m(t_2) \quad ; \quad \text{only "vertical motion" block needed.}$$

Model Parametrization

considerations:

- linear or nonlinear system?
- frequency content of excitation?
- noise at input and output?
- safety issues?

data purpose:

- identify system and parameters
- validate result of modeling
↳ don't use same dataset !!

Least Square Method for linear systems

$$y_k = h^T(u_k) \cdot \pi + e_k$$

$$\left(\begin{array}{lll} k \in [1, \dots, r] & : \text{index of measurement} & e_k \in \mathbb{R} : k\text{-th measurement error} \\ u_k \in \mathbb{R}^m & : k\text{-th input vector} & h(u_k) = h_k \in \mathbb{R}^q : \text{regressor function, nonlinear} \\ y_k \in \mathbb{R} & : k\text{-th output measurement} & \pi \in \mathbb{R}^q : \text{parameters of system, } q \leq r \end{array} \right)$$

$$\left. \begin{array}{l} \tilde{y} = [y_1, \dots, y_r]^T \in \mathbb{R}^r \\ \tilde{e} = [e_1, \dots, e_r]^T \in \mathbb{R}^r \\ H = [h_1, \dots, h_r]^T \in \mathbb{R}^{r \times q} \end{array} \right\} \begin{array}{l} \tilde{e} = \tilde{y} - H \cdot \pi \\ \hookrightarrow \text{minimize: } \epsilon = \tilde{e}^T \cdot W \cdot \tilde{e} \quad \left(W = I \text{ or } W = \text{diag}\{w_i\} \text{ if } \right. \\ \left. \hookrightarrow \pi_{LS} = [H^T W H]^{-1} \cdot H^T W \tilde{y} \quad \left(\begin{array}{l} \text{not all measurements reliable} \end{array} \right) \end{array}$$

• Iterative solution $W=I \rightarrow \pi_{LS}(r) = \left(\sum_{k=1}^r h_k \cdot h_k^T \right)^{-1} \cdot \sum_{k=1}^r h_k \cdot y_k$

$$\pi_{LS}(r+1) = \pi_{LS}(r) + \underbrace{\delta_{r+1}}_{\text{correction direction}} \cdot \underbrace{(y_{r+1} - h_{r+1}^T \cdot \pi_{LS}(r))}_{\text{prediction error}} \quad (\text{new } \pi_{LS} \text{ after adding one measurement})$$

$$\begin{array}{l|l} \pi_{LS}(0) = \text{estim.} & \delta_{r+1} = \frac{1}{1+c_{r+1}} \cdot \Omega_r \cdot h_{r+1} \\ \Omega_0 = \text{??} & \uparrow \Omega_{r+1} = \Omega_r - \frac{1}{1+c_{r+1}} \cdot \Omega_r \cdot h_{r+1} \cdot h_{r+1}^T \cdot \Omega_r \quad \left(\Omega_r = \left[\sum_{k=1}^r h_k \cdot h_k^T \right]^{-1} \right) \\ & \uparrow c_{r+1} = h_{r+1}^T \cdot \Omega_r \cdot h_{r+1} \end{array}$$

• Exponential Forgetting

$$\epsilon_{(r)} = \sum_{k=1}^r \lambda^{-k} \cdot (y_k - h_k^T \cdot \pi_{LS}(k))^2 \quad (\lambda < 1) \quad (\text{old errors get less important than new})$$

$$\left\{ \begin{array}{l} \delta_{r+1} = \frac{1}{\lambda + c_{r+1}} \cdot \Omega_r \cdot h_{r+1} \\ \Omega_{r+1} = \frac{1}{\lambda} \Omega_r \left(I - \frac{1}{\lambda + c_{r+1}} h_{r+1} h_{r+1}^T \right) \cdot \Omega_r \end{array} \right. \quad (\text{rest is the same as iterative LS})$$

• Simplified recursive LS

$$\delta_{r+1} = (\gamma \cdot h_{r+1}) / (\lambda + h_{r+1}^T \cdot h_{r+1}) \quad \left(\begin{array}{l} 0 < \gamma < 2 : \text{convergence} \\ 0 < \lambda < 1 : \text{forgetting} \end{array} \right) \quad (\text{rest is the same as iterative LS})$$

Example 1

output $y = \dot{v}$, input $u = F/m$, state $x = v$, parameter $\pi_1 = -k_0/m$; $\pi_2 = -k_1/m$

model: $m \cdot \frac{d}{dt} v(t) = -(k_0 + k_1 v(t)^2) + F(t)$

$\hookrightarrow \dot{x}(t) = \pi_1 + \pi_2 \cdot x(t)^2 + u(t)$; $y(t) = x(t)$

experiment: $u(t) = 0 \forall t$, set $x(0) = ?$



$$\begin{pmatrix} \dot{x}_1 \\ \vdots \\ \dot{x}_r \end{pmatrix} = \begin{pmatrix} 1 & x_1 \\ \vdots & \vdots \\ 1 & x_r \end{pmatrix} \cdot \begin{pmatrix} \pi_1 \\ \pi_2 \end{pmatrix} = H \cdot \pi \rightarrow \pi_{LS} = [H^T W H]^{-1} \cdot H^T W \dot{x} \quad (W=I)$$

Nonlinear LS Methods (SISO)

$\frac{d}{dt} \hat{x}(t) = f(\hat{x}(t), u(t), \hat{\pi})$ $\hat{y}(t) = h(\hat{x}(t), u(t), \hat{\pi})$ ($\hat{x} \in \mathbb{R}^n$ state, $u \in \mathbb{R}$ input, $\hat{y} \in \mathbb{R}$ output)

$k \in [1, \dots, r]$ measurements, $t_i = [t_1, \dots, t_r]^T$ time, $u_i = u(t_i)$ etc.

minimize: $\epsilon = \sum_{i=1}^r s_i \cdot (y_i(\pi) - \hat{y}_i(\hat{\pi}))^2 \rightarrow$ use comp methods!
weight real model

Analysis of linear systems

non linear: $\dot{\underline{z}}(t) = f(\underline{z}(t), \underline{v}(t), t)$, $\underline{w}(t) = g(\underline{z}(t), \underline{v}(t), t)$ state $\underline{z}, \underline{x} \in \mathbb{R}^n$
 linear: $\dot{\underline{x}}(t) = \underline{A}\underline{x}(t) + \underline{B}\underline{u}(t)$, $\underline{y}(t) = \underline{C}\underline{x}(t) + \underline{D}\underline{u}(t)$ input $\underline{v}, \underline{u} \in \mathbb{R}^m$
 output $\underline{w}, \underline{y} \in \mathbb{R}^p$

• Normalization

$\underline{x}(t) = \underline{z}(t)/\underline{z}_0 \rightarrow \dot{\underline{x}}(t) = \dot{\underline{z}}(t)/\underline{z}_0$, $\underline{u}(t) = \underline{v}(t)/\underline{v}_0$, $\underline{y}(t) = \underline{w}(t)/\underline{w}_0$ $\underline{z}_0, \underline{v}_0, \underline{w}_0$: some reference value
 $\hookrightarrow \dot{\underline{x}}(t) = \underline{f}_0(\underline{x}(t), \underline{u}(t), t)$, $\underline{y}(t) = \underline{g}_0(\underline{x}(t), \underline{u}(t), t)$

• Linearization

$\underline{f}_0(\underline{x}_e, \underline{u}_e, t) = \underline{0} \rightarrow \tilde{\underline{x}}(t) = \underline{x}(t) - \underline{x}_e$, $\tilde{\underline{u}}(t) = \underline{u}(t) - \underline{u}_e$, $\tilde{\underline{y}}(t) = \underline{y}(t) - \underline{g}_0(\underline{x}_e, \underline{u}_e, t)$
 $\hookrightarrow \dot{\tilde{\underline{x}}}(t) = \tilde{\underline{f}}_0(\tilde{\underline{x}}(t), \tilde{\underline{u}}(t), t)$, $\tilde{\underline{y}}(t) = \tilde{\underline{g}}_0(\tilde{\underline{x}}(t), \tilde{\underline{u}}(t), t)$

$$\underline{A} = \begin{pmatrix} \frac{\partial f_{0,1}}{\partial x_1} \Big|_{\substack{\underline{x}=\underline{x}_e \\ \underline{u}=\underline{u}_e}} & \dots & \frac{\partial f_{0,1}}{\partial x_n} \Big|_{\substack{\underline{x}=\underline{x}_e \\ \underline{u}=\underline{u}_e}} \\ \vdots & & \vdots \\ \frac{\partial f_{0,n}}{\partial x_1} \Big|_{\substack{\underline{x}=\underline{x}_e \\ \underline{u}=\underline{u}_e}} & \dots & \frac{\partial f_{0,n}}{\partial x_n} \Big|_{\substack{\underline{x}=\underline{x}_e \\ \underline{u}=\underline{u}_e}} \end{pmatrix}^{(n \times n)} \quad \underline{B} = \begin{pmatrix} \frac{\partial f_{0,1}}{\partial u_1} \Big|_{\substack{\underline{x}=\underline{x}_e \\ \underline{u}=\underline{u}_e}} & \dots & \frac{\partial f_{0,1}}{\partial u_m} \Big|_{\substack{\underline{x}=\underline{x}_e \\ \underline{u}=\underline{u}_e}} \\ \vdots & & \vdots \\ \frac{\partial f_{0,n}}{\partial u_1} \Big|_{\substack{\underline{x}=\underline{x}_e \\ \underline{u}=\underline{u}_e}} & \dots & \frac{\partial f_{0,n}}{\partial u_m} \Big|_{\substack{\underline{x}=\underline{x}_e \\ \underline{u}=\underline{u}_e}} \end{pmatrix}^{(n \times m)}$$

$$\underline{C} = \begin{pmatrix} \frac{\partial g_{0,1}}{\partial x_1} \Big|_{\substack{\underline{x}=\underline{x}_e \\ \underline{u}=\underline{u}_e}} & \dots & \frac{\partial g_{0,1}}{\partial x_n} \Big|_{\substack{\underline{x}=\underline{x}_e \\ \underline{u}=\underline{u}_e}} \\ \vdots & & \vdots \\ \frac{\partial g_{0,p}}{\partial x_1} \Big|_{\substack{\underline{x}=\underline{x}_e \\ \underline{u}=\underline{u}_e}} & \dots & \frac{\partial g_{0,p}}{\partial x_n} \Big|_{\substack{\underline{x}=\underline{x}_e \\ \underline{u}=\underline{u}_e}} \end{pmatrix}^{(p \times n)} \quad \underline{D} = \begin{pmatrix} \frac{\partial g_{0,1}}{\partial u_1} \Big|_{\substack{\underline{x}=\underline{x}_e \\ \underline{u}=\underline{u}_e}} & \dots & \frac{\partial g_{0,1}}{\partial u_m} \Big|_{\substack{\underline{x}=\underline{x}_e \\ \underline{u}=\underline{u}_e}} \\ \vdots & & \vdots \\ \frac{\partial g_{0,p}}{\partial u_1} \Big|_{\substack{\underline{x}=\underline{x}_e \\ \underline{u}=\underline{u}_e}} & \dots & \frac{\partial g_{0,p}}{\partial u_m} \Big|_{\substack{\underline{x}=\underline{x}_e \\ \underline{u}=\underline{u}_e}} \end{pmatrix}^{(p \times m)}$$

$\hookrightarrow \dot{\underline{x}}(t) = \underline{A}\underline{x}(t) + \underline{B}\underline{u}(t)$, $\underline{y}(t) = \underline{C}\underline{x}(t) + \underline{D}\underline{u}(t)$

• Reversion

$\underline{z}(t) = \underline{z}_0(\underline{x}(t) + \underline{x}_e)$, $\underline{v}(t) = \underline{v}_0(\underline{u}(t) + \underline{u}_e)$, $\underline{w}(t) = \underline{w}_0(\underline{y}(t) + \underline{y}_e)$

• Solution

$\underline{x}(t) = \underline{\Phi}(t)\underline{x}(0) + \int_0^t \underline{\Phi}(t-\sigma)\underline{B}\underline{u}(\sigma) d\sigma$

$\hookrightarrow \underline{\Phi}(t) = e^{\underline{A}t} = \underline{I} + \underline{A}t + (\underline{A}t)^2/2! + \dots + (\underline{A}t)^n/n! + \dots$

$\underline{y}(t) = \underline{C}\underline{x}(t) + \underline{D}\underline{u}(t)$

- **Lyapunov Stability** ($u(t)=0 \rightarrow \underline{x}(t) = \underline{\Phi} \cdot \underline{x}(0)$ finite for $t \rightarrow \infty$?)
 - asymptotically stable : $\lim_{t \rightarrow \infty} \|\underline{x}(t)\| = 0 \Leftrightarrow \forall \text{EW}(\underline{A}) < 0$
 - stable (*) (**): $\|\underline{x}(t)\| < \infty \forall t \in (0, \infty) \Leftrightarrow \forall \text{EW}(\underline{A}) \leq 0$
 - unstable : $\lim_{t \rightarrow \infty} \|\underline{x}(t)\| = \infty \Leftrightarrow \exists \text{EW}(\underline{A}) > 0$

$$\|\underline{x}(t)\| = \sqrt{\sum_{i=1}^n x_i^2(t)} \quad / \quad \text{EW}(\underline{A}) = \lambda_i \leftarrow \det(\lambda_i \mathbb{I} - \underline{A}) = 0 \quad \text{incomplete!}$$

(*) if $\exists i \neq j$ with $\lambda_i = \lambda_j$ and $\text{Re}(\lambda_i) = \text{Re}(\lambda_j) = 0$ then could be unstable! \leftarrow

(**) linear system stable \Leftrightarrow real system stable

$$\begin{pmatrix} \lambda_1 & 1 & & \\ & \lambda_1 & & \\ & & \lambda_3 & 1 \\ & & & \lambda_3 & 1 \\ & & & & \lambda_3 \end{pmatrix}$$

- **Reachability** ($\forall \underline{x}$ can be reached with some $u(t)$ from any $\underline{x}(0)$)

$$\underline{R}_n = [\underline{B}, \underline{A}\underline{B}, \underline{A}^2\underline{B}, \dots, \underline{A}^{n-1}\underline{B}] \in \mathbb{R}^{n \times n} \text{ has full Rank } n \quad (\det(\underline{R}_n) \neq 0)$$

- **Observability** (possible to uniquely reconstruct $\underline{x}(0)$ from $y(t)$ with $u(t)=0 \forall t$)

$$\underline{O}_n = [\underline{C}; \underline{C}\underline{A}; \underline{C}\underline{A}^2; \dots; \underline{C}\underline{A}^{n-1}]^T \in \mathbb{R}^{n \times n} \text{ has full Rank } n \quad (\det(\underline{O}_n) \neq 0)$$

Example 1

$$\dot{\underline{x}} = \underline{A} \cdot \underline{x} + \underline{B} \cdot u, \quad y = \underline{C} \cdot \underline{x} + \underline{D} \cdot u$$

$$\begin{pmatrix} \delta \dot{\psi} \\ \delta \ddot{\psi} \\ \delta \dot{x} \\ \delta \ddot{x} \end{pmatrix} = \begin{pmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & a_1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & a_2 & 0 \end{pmatrix} \cdot \begin{pmatrix} \delta \psi \\ \delta \dot{\psi} \\ \delta x \\ \delta \dot{x} \end{pmatrix} + \begin{pmatrix} 0 \\ b_1 \\ 0 \\ b_2 \end{pmatrix} \cdot \delta u, \quad \delta y = \begin{pmatrix} 0 \\ 0 \\ c_1 \\ 0 \end{pmatrix}^T \cdot \begin{pmatrix} \delta \psi \\ \delta \dot{\psi} \\ \delta x \\ \delta \dot{x} \end{pmatrix} + (0) \cdot \delta u$$

$$a_1 = \frac{mgR\vartheta}{\Gamma}, \quad b_1 = \frac{mr^2\vartheta}{\Gamma}, \quad c_1 = R+r$$

$$a_2 = \frac{mg(R^2\vartheta + r^2\vartheta)}{(R+r)\Gamma}, \quad b_2 = \frac{R\vartheta}{(R+r)\Gamma}$$

with some numbers for $R, r, \vartheta, \vartheta, g, m, \Gamma$:

$$\underline{R}_n = \begin{pmatrix} 0 & 26 & 0 & 1 \\ 26 & 0 & 1 & 0 \\ 0 & 7.8 & 0 & 9.1 \\ 7.8 & 0 & 9.1 & 0 \end{pmatrix} \rightarrow \det(\underline{R}_n) = 263.4 \rightarrow \text{fully controllable}$$

$$\underline{O}_n = \begin{pmatrix} 0 & 0 & 0.3 & 0 \\ 0 & 0 & 0 & 0.3 \\ 0 & 0 & 7.4 & 0 \\ 0 & 0 & 0 & 7.4 \end{pmatrix} \rightarrow \det(\underline{O}_n) = 0 \rightarrow \psi, \dot{\psi} \text{ not observable.}$$

$$\text{EW}(\underline{A}): \lambda_1 = \lambda_2 = 0, \lambda_3 = -\lambda_4 = 3.03 \rightarrow \text{unstable!}$$

Balanced Realization and Order Reduction

• Controllability/Observability measure

$$A = \begin{pmatrix} -1 & \varepsilon & 0 \\ 0 & 0 & 1 \\ 0 & a_0 & a_1 \end{pmatrix} \quad B = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \quad \begin{array}{l} \text{if } \varepsilon \neq 0 \rightarrow \det(R_n) \neq 0 \rightarrow \text{fully controllable} \\ \hookrightarrow \text{but if } \varepsilon \text{ very small} \rightarrow \text{hard to control } x_1 \end{array}$$

controllability measure: $W_R = \int_0^\infty e^{A\sigma} B B^T e^{\bar{A}\sigma} d\sigma \Leftrightarrow A W_R + W_R A = -B B^T$

$$\hookrightarrow W_R(\varepsilon=1) = \begin{pmatrix} -1.5 & -1.5 & 0.5 \\ -1.5 & -2 & 0 \\ 0.5 & 0 & -1 \end{pmatrix} \quad W_R(\varepsilon=0.01) = \begin{pmatrix} \approx 0 & \approx 0 & \approx 0 \\ \approx 0 & -2 & 0 \\ \approx 0 & 0 & -1 \end{pmatrix} \rightarrow \det(W_R) \approx 0$$

$\rightarrow x_1 \text{ hard to control}$

observability measure: $W_o = \int_0^\infty e^{\bar{A}\sigma} C^T C e^{A\sigma} d\sigma \Leftrightarrow A^T W_o + W_o A = -C^T C$

• Order Reduction

bad idea: delete system parts that do not contribute to W_R, W_o . (W_R could compensate W_o)

good idea:

find transformation $T x_b = x$ so that $W_{R,b} = W_{o,b} = \text{diag}(\sigma_i)$

\hookrightarrow relatively small σ_i can be neglected, but not their gain contribution!

$$\textcircled{1} \quad \left. \begin{array}{l} W_R = V_R \Lambda_R^2 V_R^T \rightarrow T_R = V_R \Lambda_R \\ (\tilde{W}_o = T_R^T W_o T_R) = V_o \Lambda_o^2 V_o^T \rightarrow T_o = V_o \Lambda_o^{1/2} \end{array} \right\} T = T_R T_o \rightarrow W_{R,b} = W_{o,b} = \begin{pmatrix} \sigma_1 & 0 & \dots & 0 \\ 0 & \sigma_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & \sigma_n \end{pmatrix}$$

$\textcircled{2}$ new system: $\dot{x}_b = \tilde{A} x_b + \tilde{B} u$, $y = \tilde{C} x_b + \tilde{D} u$ with $\tilde{A} = T^{-1} A T$, $\tilde{B} = T^{-1} B$, $\tilde{C} = C T$, $\tilde{D} = D$

$\textcircled{3}$ divide system: $\frac{d}{dt} \begin{pmatrix} x_{b1} \\ x_{b2} \end{pmatrix} = \begin{pmatrix} \tilde{A}_{11} & \tilde{A}_{12} \\ \tilde{A}_{21} & \tilde{A}_{22} \end{pmatrix} \begin{pmatrix} x_{b1} \\ x_{b2} \end{pmatrix} + \begin{pmatrix} \tilde{B}_1 \\ \tilde{B}_2 \end{pmatrix} u$ $y = (\tilde{C}_1 \mid \tilde{C}_2) \begin{pmatrix} x_{b1} \\ x_{b2} \end{pmatrix} + \tilde{D} u$

\uparrow where x_{b2} = system parts that can be neglected (low σ_i !)

$\textcircled{4}$ new system: $\dot{x}_{b1} = \tilde{A}_{11} x_{b1} + \tilde{B}_1 u$, $y = \tilde{C}_1 x_{b1} + \tilde{D} u$

to match gain: $\tilde{A}_{11} \rightarrow \tilde{A}_{11} - A_{12} A_{22}^{-1} A_{21}$ $\tilde{C}_1 \rightarrow C_1 - C_2 A_{22}^{-1} A_{21}$ (singular perturbation)
 $\tilde{B}_1 \rightarrow \tilde{B}_1 - A_{12} A_{22}^{-1} B_2$ $\tilde{D} \rightarrow \tilde{D} - C_2 A_{22}^{-1} B_2$

Zero Dynamics (SISO)

$$\dot{x} = Ax + Bu \quad y = Cx + Du$$

$$P(s) = C(sI - A)^{-1}B + D \rightarrow P(s) = k \cdot \frac{1 \cdot s^{n-r} + b_{n-r-1}s^{n-r-1} + \dots + b_1s + b_0}{1 \cdot s^n + a_{n-1}s^{n-1} + \dots + a_1s + a_0}$$

(n : system order)
(k : input gain)
(r : relative degree)

$$\dot{x} = \begin{pmatrix} 0 & 1 & 0 & \dots & 0 \\ 0 & 0 & 1 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & 1 \\ -a_0 & -a_1 & -a_2 & \dots & -a_{n-1} \end{pmatrix} \cdot x + \begin{pmatrix} 0 \\ 0 \\ 0 \\ \vdots \\ k \end{pmatrix} \cdot u \quad y = \begin{pmatrix} b_0 \\ \vdots \\ b_{n-r-1} \\ 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix} \cdot x$$

Zero Dynamics: $u(t) \neq 0$ and $x(0) = ? \rightarrow y(t) = 0$ for finite $t \rightarrow$ problematic for controller.

$$\begin{aligned} z_1 = y &= Cx &= b_0x_1 + b_1x_2 + \dots + b_{n-r-1}x_{n-r} + x_{n-r+1} \\ z_2 = \dot{y} &= C\dot{x} = CAx + CBx = CAx &= b_0x_2 + b_1x_3 + \dots + b_{n-r-1}x_{n-r+1} + x_{n-r+2} \\ z_3 = \ddot{y} &= CA\dot{x} = CA^2x + CABx = CA^2x &= b_0x_3 + b_1x_4 + \dots + b_{n-r-1}x_{n-r+2} + x_{n-r+3} \\ &\vdots &\vdots \\ z_r = y^{(r-1)} &= CA^{r-1}x + CA^{r-2}Bu = CA^{r-1}x &= b_0x_r + b_1x_{r+1} + \dots + b_{n-r-1}x_{n-1} + x_n \\ \left(y^{(r)} = CA^r x + CA^{r-1}Bu = CA^r x + \underline{k}u \right. &= b_0x_{r+1} + b_1x_{r+2} + \dots + b_{n-r-1}x_n + \underline{\dot{x}_n} \end{aligned}$$

$$z_{r+1} = x_1$$

$$z_{r+2} = x_2$$

$$\vdots$$

$$z_n = x_{n-r}$$

$$z = \Phi x = \begin{pmatrix} \xi = \begin{pmatrix} z_1 \\ \vdots \\ z_r \end{pmatrix} \\ \eta = \begin{pmatrix} z_{r+1} \\ \vdots \\ z_n \end{pmatrix} \end{pmatrix}$$

(coordinate transform)

$$\begin{pmatrix} \dot{\xi} \\ \dot{\eta} \end{pmatrix} = \left(\begin{array}{c|c} \begin{pmatrix} 0 & 1 & 0 & \dots & 0 \\ 0 & 0 & 1 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & 1 \end{pmatrix} & \begin{pmatrix} 0 & 0 & 0 & \dots & 0 \\ 0 & 0 & 0 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & 0 \end{pmatrix} \\ \hline \begin{pmatrix} 0 & 0 & 0 & \dots & 0 \\ 0 & 0 & 0 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & 0 \end{pmatrix} & \begin{pmatrix} 0 & 1 & 0 & \dots & 0 \\ 0 & 0 & 1 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & 1 \end{pmatrix} \end{array} \right) \cdot \begin{pmatrix} \xi \\ \eta \end{pmatrix} + \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ k \\ 0 \\ \vdots \\ 0 \end{pmatrix} \cdot u, \quad y = z_1 = \xi(1)$$

$$\left(\begin{array}{c|c} R & S \\ \hline P & Q \end{array} \right)$$

to have a vanishing output: $\xi^*(0) = 0$, $\dot{u}(t) = -\frac{1}{k} S^T \eta^*(t)$ ($\eta^*(0)$ arbitrary)

$\hookrightarrow y(t) = 0$ and $\xi(t) = 0 \forall t \rightarrow$ zero dynamics: $\dot{\eta}^*(t) = Q \cdot \eta^*(t)$

$$Q = \begin{pmatrix} 0 & 1 & 0 & \dots & 0 \\ 0 & 0 & 1 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & 1 \\ \hline & & & & q^T \end{pmatrix} \leftarrow q = \begin{pmatrix} -b_0 \\ -b_1 \\ \vdots \\ -b_{n-r-2} \\ -b_{n-r-1} \end{pmatrix}$$

Q asympt. stable \rightarrow system is minimum phase $\leftarrow \forall \operatorname{Re}(\text{zero}) < 0$

Example: ($n=4, r=2$)

$$P(s) = \frac{Y(s)}{U(s)} = k \frac{b_1 s + b_0}{a_3 s^3 + a_2 s^2 + a_1 s + a_0} \Rightarrow \dot{x}(t) = \begin{pmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ -a_0 & -a_1 & -a_2 & -a_3 \end{pmatrix} \cdot x(t) + \begin{pmatrix} 0 \\ 0 \\ 0 \\ k \end{pmatrix} \cdot u(t)$$

$$y(t) = (b_0 \ b_1 \ 1 \ 0) \cdot x(t) + (0) \cdot u(t)$$

$$\left. \begin{aligned} z_1 &= y = b_0 x_1 + b_1 x_2 + x_3 \\ z_2 &= \dot{y} = b_0 x_2 + b_1 x_3 + x_4 \\ \ddot{y} &= -a_0 x_1 - a_1 x_2 + (b_0 - a_2) x_3 \\ &\quad + (b_1 - a_3) x_4 + k u \end{aligned} \right\}$$

$$z_3 = x_1$$

$$z_4 = x_2$$

$$z = \Phi^{-1} x$$

$$x = \Phi z$$

$$\downarrow \Phi^{-1} = \begin{pmatrix} b_0 & b_1 & 1 & 0 \\ 0 & b_0 & b_1 & 1 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{pmatrix} \rightarrow \Phi = \begin{pmatrix} 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 1 & 0 & -b_0 & -b_1 \\ -b_1 & 1 & b_0 b_1 & b_1^2 b_0 \end{pmatrix}$$

$$\Rightarrow \dot{z} = \Phi^{-1} A \Phi z + \Phi^{-1} B u, \quad y = C \Phi z$$

$$\frac{d}{dt} \begin{pmatrix} \xi_1 \\ \xi_2 \\ \eta_1 \\ \eta_2 \end{pmatrix} = \begin{pmatrix} 0 & 1 & 0 & 0 \\ r_1 & r_2 & s_1 & s_2 \\ 0 & 0 & 0 & 1 \\ 1 & 0 & -b_0 & -b_1 \end{pmatrix} \cdot \begin{pmatrix} \xi_1 \\ \xi_2 \\ \eta_1 \\ \eta_2 \end{pmatrix} + \begin{pmatrix} 0 \\ k \\ 0 \\ 0 \end{pmatrix} \cdot u \leftarrow \begin{cases} r_1 = b_0 - a_2 - b_1(b_1 - a_3) \\ r_2 = b_1 - a_3 \\ s_1 = b_0 b_1(b_1 - a_3) - a_0 - b_0(b_0 - a_2) \\ s_2 = (b_1 - a_3)(b_1^2 - b_0) - a_1 - (b_0 - a_2)b_1 \end{cases}$$

Case Study: Geostationary Satellite

Lagrange Method: $L = T - V$

$$T = \frac{1}{2} m \dot{r}^2 + \frac{1}{2} m (r \dot{\varphi})^2$$

$$V = \int_R^r F(s) ds = \int_R^r G \frac{M \cdot m}{s^2} ds = GMm \left(\frac{1}{R} - \frac{1}{r} \right)$$

$$\frac{d}{dt} \left[\frac{\partial L}{\partial \dot{r}} \right] - \frac{\partial L}{\partial r} = F_r$$

$$\frac{d}{dt} \left[\frac{\partial L}{\partial \dot{\varphi}} \right] - \frac{\partial L}{\partial \varphi} = F_\varphi \cdot r$$

$$\left(\begin{array}{ll} \frac{\partial L}{\partial \dot{r}} = m \dot{r} & \frac{\partial L}{\partial \dot{\varphi}} = m r^2 \dot{\varphi} \\ \frac{d}{dt} \left[\frac{\partial L}{\partial \dot{r}} \right] = m \ddot{r} & \frac{d}{dt} \left[\frac{\partial L}{\partial \dot{\varphi}} \right] = m r^2 \ddot{\varphi} + 2 m r \dot{\varphi} \\ \frac{\partial L}{\partial r} = m r \dot{\varphi}^2 - GMm \frac{1}{r^2} & \frac{\partial L}{\partial \varphi} = 0 \end{array} \right)$$

$$\rightarrow m \ddot{r} = m r \dot{\varphi}^2 - GMm \frac{1}{r^2} + F_r$$

$$\rightarrow m r^2 \ddot{\varphi} = -2 m r \dot{\varphi} \dot{r} + F_\varphi r$$

inputs: $u_1 = F_r/m$; $u_2 = F_\varphi/m$

equilibrium: $u_1 = 0$, $\ddot{r} = 0$, $\dot{r} = 0$, $r = r_0$

$$u_2 = 0, \ddot{\varphi} = 0, \dot{\varphi} = \omega_0, \varphi = \omega_0 t$$

$$\omega_0 = 7.29 \cdot 10^{-5} \text{ rad/s} \quad \text{geostationary orbit}$$

$$\rightarrow r_0 \omega_0^2 - GM \frac{1}{r_0^2} = 0 \rightarrow r_0 = \left(\frac{GM}{\omega_0^2} \right)^{1/3}$$

states: $x_1 = r$, $x_2 = \dot{r}$, $x_3 = \varphi$, $x_4 = \dot{\varphi}$

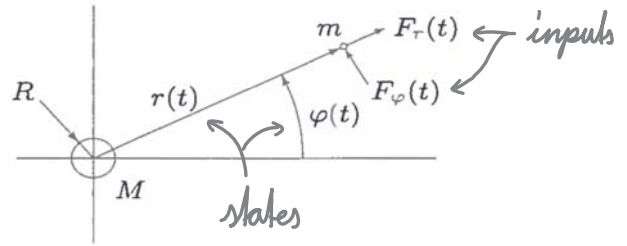
$$\frac{d}{dt} \underline{x}(t) = \underline{f}(\underline{x}(t), \underline{u}(t))$$

$$\begin{pmatrix} \dot{x}_1 \\ \dot{x}_2 \\ \dot{x}_3 \\ \dot{x}_4 \end{pmatrix} = \begin{pmatrix} x_2 \\ x_1 x_4^2 - GM/x_1^2 + u_1 \\ x_4 \\ -2x_2 x_4/x_1 + u_2/x_1 \end{pmatrix}$$

outputs: $y_1 = x_1/r_0$; $y_2 = x_3$

$$\underline{y}(t) = \underline{h}(\underline{x}(t), \underline{u}(t))$$

$$\begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \begin{pmatrix} x_1/r_0 \\ x_3 \end{pmatrix}$$



$$\underline{x}_0(t) = \begin{pmatrix} r_0 \\ 0 \\ \omega_0 t \\ \omega_0 \end{pmatrix}, \quad \underline{u}_0(t) = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$\underline{A} = \begin{pmatrix} 0 & 1 & 0 & 0 \\ 3\omega_0^2 & 0 & 0 & 2r_0\omega_0 \\ 0 & 0 & 0 & 1 \\ 0 & -2\frac{\omega_0}{r_0} & 0 & 0 \end{pmatrix}, \quad \underline{B} = \begin{pmatrix} 0 & 0 \\ 1 & 0 \\ 0 & 0 \\ 0 & 1/r_0 \end{pmatrix}$$

$$\underline{C} = \begin{pmatrix} 1/r_0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{pmatrix}, \quad \underline{D} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$$

stability: $\det(sI - A) = s^2(s^2 + \omega_0^2)$

$$\rightarrow \lambda_{1,2} = 0, \lambda_{3,4} = \pm i\omega_0 \rightarrow \text{unstable?}$$

controllability: $R_n = [B, AB, A^2B, \dots]$ has full rank $\rightarrow \det(R_n) \neq 0$

$$R_n = \begin{pmatrix} 0 & 0 & 1 & 0 & \dots \\ 1 & 0 & 0 & 2\omega_0 & \dots \\ 0 & 0 & 0 & \frac{1}{r_0} & \dots \\ 0 & \frac{1}{r_0} & -\frac{2\omega_0}{r_0} & 0 & \dots \end{pmatrix} \rightarrow \text{controllable}$$

observability: $O_n = [C; CA; CA^2, \dots]$ has full rank $\rightarrow \det(O_n) \neq 0$

$$O_n = \begin{pmatrix} \frac{1}{r_0} & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & \frac{1}{r_0} & 0 & 0 \\ 0 & 0 & 0 & 1 \\ \vdots & \vdots & \vdots & \vdots \end{pmatrix} \rightarrow \text{observable}$$

transfer function: $P(s) = \begin{pmatrix} P_{11}(s) & P_{12}(s) \\ P_{21}(s) & P_{22}(s) \end{pmatrix} = C(sI-A)^{-1}B = \frac{C \cdot \text{Adj}(sI-A)B}{\det(sI-A)}$

$$P(s) = \begin{pmatrix} \frac{1}{r_0(s^2+\omega_0^2)} & \frac{2\omega_0}{r_0s(s^2+\omega_0^2)} \\ \frac{-2}{r_0s(s^2+\omega_0^2)} & \frac{s^2-3\omega_0^2}{r_0s^2(s^2+\omega_0^2)} \end{pmatrix}$$